

On the variational formulation of some Mean field game systems

Daniela Tonon

CEREMADE, Université Paris-Dauphine, France

PGMO DAYS, November 9th 2016



Mean field games (MFG) theory analyzes :

- optimal control problems
- with (infinitely) many identical controllers

In other words it is a mathematical modeling approach to continuous-time systems which involve a large number of agents

MFG were introduced by [Lasry and Lions](#) and by [Huang, Caines and Malhamé](#) in 2006 to describe Nash equilibria in differential games with infinitely many players

Typical features of the model :

- players act according to the same principles (they are indistinguishable and have the same optimization criteria)
- players have individually a minor (infinitesimal) influence, but their strategy takes into account the mass of co-players

Roughly : players are particles but have strategies (more sophisticated than particle physics or similar economics models); however, they only consider the statistical state of the mass of co-players (less sophisticated than agents of N-players games)

Goal : introduce a macroscopic description through a mean field approach as the number of players $N \rightarrow +\infty$

Motivations :

- **Problems arising in economy :**
 - financial markets (Price formation and dynamic equilibria, Formation of volatility) (Lasry, Lions, 2006)
 - general economic equilibrium with rational expectations (Guéant, Lasry, and Lions, 2007)
- **Dynamics of population models :**
 - crowd motion : mexican wave "la ola", ... (Guéant, Lasry, Lions - Lachapelle, ...)
 - academic behavior (Besancenot, Courtault, El Dika...)
- **Engineering literature :** Large Population Stochastic Wireless Power Control Problem (Huang, Caines, Malhamé, 2003, Mériaux, Lasaulce...)

Different approaches

- **limit of N -player (stochastic) differential games as $N \rightarrow +\infty$**
→ analogy with the Mean Field theories in statistical physics (kinetic theory of gases, Boltzmann and Vlasov equations) and quantum mechanics and quantum Chemistry (Hartree-Fock models...)
- **direct definition** of (stochastic) differential games with infinitely many identical players (applications to N -player games),
→ approach from game theory
- **potential games** : games arising as necessary conditions for optimal control problems of PDE equations
→ related to optimal transportation problems

The 3 approaches yield the same MFG system with unknown (ϕ, m) written here in a second order form with degenerate diffusion and a local coupling

$$\begin{cases} -\partial_t \phi - A_{ij} \partial_{ij} \phi + H(x, D\phi) = f(x, m(x, t)) \\ \partial_t m - \partial_{ij}(A_{ij} m) - \operatorname{div}(m D_p H(x, D\phi)) = 0 \\ m(0) = m_0, \phi(x, T) = \phi_T(x) \end{cases}$$

where

- $A : \mathbb{R}^d \rightarrow \mathbb{R}^{d \times d}$ is symmetric and nonnegative
- the Hamiltonian $H : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$ is convex in the second variable
- the coupling $f : \mathbb{R}^d \times [0, +\infty) \rightarrow [0, +\infty)$ is increasing with respect to the second variable
- m_0 is a probability density
- $\phi_T : \mathbb{R}^d \rightarrow \mathbb{R}$ is a given function

The optimal control problem of a typical small player

$\phi(t, x)$ is the **value function**

Dynamics :

$$dX_s = v_s ds + \Sigma(X_s) dB_s,$$

where (v_s) is the control, (B_s) is a Brownian motion and $\Sigma \Sigma^T = A$

Cost :

$$\mathbb{E} \left[\int_0^T H^*(X_s, -v_s) + f(X_s, m(s, X_s)) ds + \phi_T(X_T) \right]$$

$\forall t \in [0, T]$ $m(t, x)$ denotes the **density** of population of small players at position x

the **optimal control** is formally given by the feedback
 $(t, x) \rightarrow -D_p H(x, D\phi(t, x))$

the second equation in (MFG) is the **Kolmogorov equation** of the process (X_s) when the small player plays in an optimal way

In the MFG systems with **uniformly parabolic diffusions** (typically $A_{ij}\partial_{ij}\phi = \Delta\phi$) the solutions are **smooth**, (at least if the coupling is nonlocal and regularizing or if it has a "small growth")

Analysis by **PDE methods** : see Cardaliaguet, Lasry, Lions and Porretta (2012), Lasry and Lions (2006, 2007), Gomes, Pimentel and Sánchez-Morgado (2013), Lions (2011), Porretta (2013)

Analysis by **stochastic techniques** : see Carmona and Delarue (2013), Huang, Malhamé and Caines (2006)

The case of **couplings with an arbitrary growth** has been discussed in Cardaliaguet, Lasry, Lions and Porretta (2012) (for quadratic hamiltonians and smooth solutions) and in Porretta (2013) (for more general hamiltonians but weak solutions)

Since we consider **degenerate parabolic equations** we face a lack of regularity \Rightarrow break down of the uniformly parabolic techniques

IDEA : use convex optimization methods from optimal transport problems (see Benamou and Brenier (2000), Cardaliaguet, Carlier and Nazaret (2012))

These techniques were already used to study first order MFG systems (i.e., $A \equiv 0$): see Cardaliaguet (2013), Cardaliaguet and Graber (2013), Graber (2013)

We show the existence and uniqueness of a weak solution for the degenerate MFG system

Assumptions

- (H1) $f : \mathbb{T}^d \times [0, +\infty) \rightarrow \mathbb{R}$ is continuous in both variables, **increasing** wrt the second variable m , and $\exists q > 1, C_1$ s.t.

$$\frac{1}{C_1}|m|^{q-1} - C_1 \leq f(x, m) \leq C_1|m|^{q-1} + C_1 \quad \forall m \geq 0$$

Moreover $f(x, 0) = 0 \quad \forall x \in \mathbb{T}^d$

- (H2) $H : \mathbb{T}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$ is continuous in both variables, **convex** and differentiable in the second variable, with $D_p H$ continuous in both variable, and $\exists r > 1, C_2 > 0$ s.t.

$$\frac{1}{rC_2}|\xi|^r - C_2 \leq H(x, \xi) \leq \frac{C_2}{r}|\xi|^r + C_2 \quad \forall (x, \xi) \in \mathbb{T}^d \times \mathbb{R}^d$$

Assumptions

- (H3) $A : \mathbb{T}^d \rightarrow \mathbb{R}^{d \times d}$ has **symmetric nonnegative** values and there exists $\Sigma : \mathbb{T}^d \rightarrow \mathbb{R}^{d \times d}$ Lipschitz continuous such that $\Sigma \Sigma^T = A$:
 $\exists C_3 > 0$ s.t.

$$|\Sigma(x) - \Sigma(y)| \leq C_3 |x - y| \quad \forall x, y \in \mathbb{T}^d, \xi \in \mathbb{R}^d$$

Moreover, either $r \geq p$ or $A \equiv 0$, where $\frac{1}{p} + \frac{1}{q} = 1$

- (H4) $\phi_T : \mathbb{T}^d \rightarrow \mathbb{R}$ is of class C^2 , while $m_0 : \mathbb{T}^d \rightarrow \mathbb{R}$ is a C^1 positive density

The **Fenchel conjugate** H^* of H wrt the second variable

$$H^*(x, w) = \sup_{p \in \mathbb{R}^d} \langle w, p \rangle - H(x, p)$$

is continuous and

$$\frac{1}{r' C_2} |\xi|^{r'} - C_2 \leq H^*(x, \xi) \leq \frac{C_2}{r'} |\xi|^{r'} + C_2 \quad \forall (x, \xi) \in \mathbb{T}^d \times \mathbb{R}^d,$$

where r' is s.t. : $\frac{1}{r} + \frac{1}{r'} = 1$

Let

$$F(x, m) = \begin{cases} \int_0^m f(x, \tau) d\tau & \text{if } m \geq 0 \\ +\infty & \text{otherwise} \end{cases}$$

Then F is continuous on $\mathbb{T}^d \times (0, +\infty)$, derivable and **strictly convex** in m and

$$\frac{1}{qC_1}|m|^q - C_1 \leq F(x, m) \leq \frac{C_1}{q}|m|^q + C_1 \quad \forall m \geq 0$$

Let F^* be the **Fenchel conjugate** of F wrt the second variable then $F^*(x, a) = 0$ for $a \leq 0$, moreover,

$$\frac{1}{pC_1}|a|^p - C_1 \leq F^*(x, a) \leq \frac{C_1}{p}|a|^p + C_1 \quad \forall a \geq 0,$$

where p is s.t. : $1/p + 1/q = 1$

Optimal control problems in duality

An optimal control problem for a **backward HJ equation** :

the state variable ϕ is controlled by a distributed control $\alpha : (0, T) \times \mathbb{T}^d \rightarrow \mathbb{R}$ in order to minimize the criterium

$$\int_0^T \int_{\mathbb{T}^d} F^*(x, \alpha(t, x)) \, dx dt - \int_{\mathbb{T}^d} \phi(0, x) dm_0(x)$$

when the state ϕ is driven by

$$\begin{cases} -\partial_t \phi - A_{ij}(x) \partial_{ij} \phi + H(x, D\phi) = \alpha(t, x) \\ \phi(x, T) = \phi_T(x) \end{cases}$$

Precisely :

$$\inf_{\phi \in \mathcal{K}_0} \mathcal{A}(\phi)$$

where $\mathcal{K}_0 := \{\phi \in \mathcal{C}^2([0, T] \times \mathbb{T}^d) \mid \phi(T, x) = \phi_T(x)\}$ and

$$\begin{aligned} \mathcal{A}(\phi) &= \int_0^T \int_{\mathbb{T}^d} F^*(x, -\partial_t \phi(t, x) - A_{ij} \partial_{ij} \phi + H(x, D\phi(t, x))) \, dx dt \\ &\quad - \int_{\mathbb{T}^d} \phi(0, x) dm_0(x) \end{aligned}$$

The second optimal control problem is an optimal control problem for a **Fokker-Plank equation** :

we control the state variable m through a vector field $v : (0, T) \times \mathbb{T}^d \rightarrow \mathbb{R}^d$ in order to minimize

$$\int_0^T \int_{\mathbb{T}^d} m(t, x) H^*(x, -v(t, x)) + F(x, m(t, x)) \, dx dt + \int_{\mathbb{T}^d} \phi_T(x) m(T, x) \, dx$$

when m solves the Fokker-Plank equation

$$\partial_t m - \partial_{ij}(A_{ij}(x)m) + \operatorname{div}(mv) = 0$$

with $m(0) = m_0$

Precisely : $\inf_{(m,w) \in \mathcal{K}_1} \mathcal{B}(m, w)$

where

$$\mathcal{K}_1 := \left\{ (m, w) \in L^1((0, T) \times \mathbb{T}^d) \times L^1((0, T) \times \mathbb{T}^d, \mathbb{R}^d) \mid \right. \\ \left. m(t, x) \geq 0 \text{ a.e.}, \int_{\mathbb{T}^d} m(t, x) dx = 1 \text{ for a.e. } t \in (0, T), \text{ and} \right. \\ \left. \partial_t m - \partial_{ij}(A_{ij}(x)m) + \operatorname{div}(w) = 0 \text{ in } (0, T) \times \mathbb{T}^d, \quad m(0) = m_0 \right. \\ \left. \text{in the sense of distribution} \right\}$$

i.e. for any test function $\zeta \in C_c^\infty([0, T] \times \mathbb{T}^d)$,

$$- \int_{\mathbb{T}^d} \zeta(0) m_0 - \int_0^T \int_{\mathbb{T}^d} m \partial_t \zeta + A_{ij} m \partial_{ij} \zeta + \langle w, \nabla \zeta \rangle = 0$$

Precisely :

$$\inf_{(m,w) \in \mathcal{K}_1} \mathcal{B}(m, w)$$

where

$$\mathcal{K}_1 := \left\{ (m, w) \in L^1((0, T) \times \mathbb{T}^d) \times L^1((0, T) \times \mathbb{T}^d, \mathbb{R}^d) \mid \right. \\ \left. m(t, x) \geq 0 \text{ a.e.}, \int_{\mathbb{T}^d} m(t, x) dx = 1 \text{ for a.e. } t \in (0, T), \text{ and} \right. \\ \left. \partial_t m - \partial_{ij}(A_{ij}(x)m) + \operatorname{div}(w) = 0 \text{ in } (0, T) \times \mathbb{T}^d, \quad m(0) = m_0 \right. \\ \left. \text{in the sense of distribution} \right\}$$

$$\mathcal{B}(m, w) = \int_0^T \int_{\mathbb{T}^d} m(t, x) H^* \left(x, -\frac{w(t, x)}{m(t, x)} \right) + F(x, m(t, x)) \, dx dt \\ + \int_{\mathbb{T}^d} \phi_T(x) m(T, x) dx$$

$$\text{for } m(t, x) = 0, \quad m(t, x) H^* \left(x, -\frac{w(t, x)}{m(t, x)} \right) = \begin{cases} +\infty & \text{if } w(t, x) \neq 0 \\ 0 & \text{if } w(t, x) = 0 \end{cases}$$

Lemma (CGPT 15)

We have

$$\inf_{\phi \in \mathcal{K}_0} \mathcal{A}(\phi) = - \min_{(m, w) \in \mathcal{K}_1} \mathcal{B}(m, w)$$

Moreover, the minimum in the right-hand side is achieved by a **unique** pair $(\bar{m}, \bar{w}) \in \mathcal{K}_1$ satisfying

$$(\bar{m}, \bar{w}) \in L^q((0, T) \times \mathbb{T}^d) \times L^{\frac{r'q}{r'+q-1}}((0, T) \times \mathbb{T}^d)$$

IDEA : Use the Fenchel-Rockafellar duality Theorem

Relaxation

In general, we do not expect problem

$$\inf_{\phi \in \mathcal{K}_0} \mathcal{A}(\phi)$$

to have a solution

IDEA : we consider an equivalent relaxed problem and see that it has a solution

Let $\gamma = \frac{rp(1+d)}{d-r(p-1)}$ if $p < 1 + \frac{d}{r}$ and $\gamma = +\infty$ if $p > 1 + \frac{d}{r}$

$$\mathcal{K} := \left\{ (\phi, \alpha) \in L^\gamma((0, T) \times \mathbb{T}^d) \times L^p((0, T) \times \mathbb{T}^d) \mid D\phi \in L^r((0, T) \times \mathbb{T}^d) \right.$$

and which satisfies in the sense of distribution

$$\left. \begin{aligned} -\partial_t \phi - A_{ij}(x) \partial_{ij} \phi + H(x, D\phi) \leq \alpha, \quad \phi(T, \cdot) \leq \phi_T \end{aligned} \right\}$$

Here "in the sense of distribution" means that for any nonnegative test function $\zeta \in C_c^\infty((0, T] \times \mathbb{T}^d)$,

$$- \int_{\mathbb{T}^d} \zeta(T) \phi_T + \int_0^T \int_{\mathbb{T}^d} \phi \partial_t \zeta + \langle D\zeta, AD\phi \rangle + \zeta(\partial_i A_{ij} \partial_j \phi + H(x, D\phi)) \leq \int_0^T \int_{\mathbb{T}^d} \alpha \zeta$$

Due to the presence of second order derivatives we do not expect the function ϕ to be BV (as in the first order case)

What about **trace properties**? ϕ has a trace in a weak sense

We extend the functional \mathcal{A} to \mathcal{K}

$$\mathcal{A}(\phi, \alpha) = \int_0^T \int_{\mathbb{T}^d} F^*(x, \alpha(x, t)) \, dx dt - \int_{\mathbb{T}^d} \phi(x, 0) m_0(x) \, dx \quad \forall (\phi, \alpha) \in \mathcal{K}$$

Proposition (CGPT 15)

We have

$$\inf_{\phi \in \mathcal{K}_0} \mathcal{A}(\phi) = \inf_{(\phi, \alpha) \in \mathcal{K}} \mathcal{A}(\phi, \alpha)$$

Inequality $\inf_{\phi \in \mathcal{K}_0} \mathcal{A}(\phi) \geq \inf_{(\phi, \alpha) \in \mathcal{K}} \mathcal{A}(\phi, \alpha)$ holds obviously

Let $(\phi, \alpha) \in \mathcal{K}$, then

$$\mathcal{A}(\phi, \alpha) \geq - \inf_{(m, w) \in \mathcal{K}_1} \mathcal{B}(m, w) = \inf_{\phi \in \mathcal{K}_0} \mathcal{A}(\phi)$$

Proposition (CGPT 15)

The relaxed problem has *at least one solution* $(\bar{\phi}, \bar{\alpha}) \in \mathcal{K}$ which is bounded below by a constant depending on $\|\phi_T\|_{C^2}$, on $\|A_{ij}\|_{C^0}$ and on $\|H(\cdot, D\phi_T)\|_{\infty}$.

Therefore, thanks to the equivalence, we have a infimum for \mathcal{A} over \mathcal{K}_0 and a minimum for \mathcal{B} over \mathcal{K}_1

$$\min_{(\phi, \alpha) \in \mathcal{K}} \mathcal{A}(\phi, \alpha) = \inf_{\phi \in \mathcal{K}_0} \mathcal{A}(\phi) = - \min_{(m, w) \in \mathcal{K}_1} \mathcal{B}(m, w)$$

Existence and uniqueness

We say that a pair $(\phi, m) \in L^\gamma((0, T) \times \mathbb{T}^d) \times L^q((0, T) \times \mathbb{T}^d)$ is a **weak solution to (MFG)** if

- (i) $D\phi \in L^r$, $mH^*(\cdot, D_p H(\cdot, D\phi)) \in L^1$ and $mD_p H(\cdot, D\phi) \in L^1$
- (ii) The following inequality holds in the sense of distribution

$$-\partial_t \phi - \partial_i (A_{ij}(x) \partial_j \phi) + (\partial_i A_{ij}) \partial_j \phi + H(x, D\phi) \leq f(x, m) \quad \text{in } (0, T) \times \mathbb{T}^d,$$

with $\phi(T, \cdot) \leq \phi_T$

- (iii) The following equation holds in the sense of distribution

$$\partial_t m - \partial_{ij} (A_{ij}(x) m) - \operatorname{div}(m D_p H(x, D\phi)) = 0 \quad \text{in } (0, T) \times \mathbb{T}^d, \quad m(0) = m_0$$

- (iv) The following equality holds :

$$\begin{aligned} & \int_0^T \int_{\mathbb{T}^d} m(t, x) (f(x, m(t, x)) + H^*(x, D_p H(x, D\phi)(t, x))) \, dx dt \\ & \quad + \int_{\mathbb{T}^d} m(T, x) \phi_T(x) - m_0(x) \phi(0, x) \, dx = 0 \end{aligned}$$

Theorem (CGPT 15)

There *exists* a weak solution $(\bar{\phi}, \bar{m})$ to (MFG).

Moreover this solution is *unique* in the following sense: if $(\bar{\phi}, \bar{m})$ and $(\bar{\phi}', \bar{m}')$ are two solutions, then $\bar{m} = \bar{m}'$ a.e. and $\bar{\phi} = \bar{\phi}'$ in $\{\bar{m} > 0\}$.

Finally, there exists a solution which is *bounded below* by a constant depending on $\|\phi_T\|_{C^2}$, on $\|A_{ij}\|_{C^0}$ and on $\|H(\cdot, D\phi_T)\|_{\infty}$.

Idea of the Proof :

Let $(\bar{m}, \bar{w}) \in \mathcal{K}_1$ be a minimizer of $\min_{(m,w) \in \mathcal{K}_1} \mathcal{B}(m, w)$
 and $(\bar{\phi}, \bar{\alpha}) \in \mathcal{K}$ be a minimizer of $\min_{(\phi, \alpha) \in \mathcal{K}} \mathcal{A}(\phi, \alpha)$

Then $\bar{w} = -\bar{m}D_p H(\cdot, D\bar{\phi})$ a.e. and $\bar{\alpha}(t, x) = f(x, \bar{m}(t, x))$ a.e.
 and $(\bar{\phi}, \bar{m})$ is a weak solution

Conversely, we can show that any weak solution $(\bar{\phi}, \bar{m})$ of (MFG) is such that :

the pair $(\bar{m}, -\bar{m}D_p H(\cdot, D\bar{\phi}))$ is the minimizer of $\min_{(m,w) \in \mathcal{K}_1} \mathcal{B}(m, w)$

while $(\bar{\phi}, f(\cdot, \bar{m}))$ is a minimizer of $\min_{(\phi, \alpha) \in \mathcal{K}} \mathcal{A}(\phi, \alpha)$

Uniqueness : Since the solution of $\min_{(m,w) \in \mathcal{K}_1} \mathcal{B}(m, w)$ is unique \bar{m} is unique

The proof of the uniqueness of $\bar{\phi}$ is more technical

The planning problem

What happens if we prescribe the terminal distribution of the agents $m(T)$?

Suggested by P.L. Lions, this is an optimal planning problem: can we drive in time T the density of the agents from the initial configuration m_0 to a target configuration m_T in a way which is optimal for the agents' cost ?

⇒ **Optimal transport problem**

In this case, the MFG system

$$\begin{cases} -\partial_t \phi - A_{ij} \partial_{ij} \phi + H(x, D\phi) = f(x, m(x, t)) \\ \partial_t m - \partial_{ij} (A_{ij} m) - \operatorname{div}(m D_p H(x, D\phi)) = 0 \end{cases}$$

is complemented with initial and terminal conditions for the density m :

$$m(0) = m_0, \quad m(T) = m_T$$

The final pay-off $\phi(T)$ is not prescribed here; it is a degree of freedom to be used to reach the target

The optimal control problem of a typical small player

This can be seen as an **optimal transport problem** for the probability densities of the process law

Dynamics :

$$dX_s = v_s ds + \Sigma(X_s) dB_s,$$
$$m_0 = \mathcal{L}(X_0), \quad m_T = \mathcal{L}(X_T)$$

minimizing :

$$\mathbb{E} \left[\int_0^T H^*(X_s, -v_s) + f(X_s, m(s, X_s)) ds \right]$$

among admissible feedback v_s

Alternatively, the pb. can be seen as an **exact controllability problem** for the Fokker-Planck equation (+ optimization among admissible drifts v):

Minimize

$$\int_0^T \int_{\mathbb{T}^d} m(t, x) H^*(x, -v(t, x)) + F(x, m(t, x)) \, dx dt$$

when m solves the Fokker-Plank equation

$$\partial_t m - \partial_{ij}(A_{ij}(x)m) + \operatorname{div}(mv) = 0$$

with $m(0) = m_0$ and $m(T) = m_T$

- Existing results :

- Lions : existence and uniqueness of smooth solutions :
 - for the deterministic case ($A_{ij} = 0$), H is a smooth superlinear and strictly convex Hamiltonian, $f(m)$ nondecreasing and bounded, m_0, m_T smooth and positive
 - for the model case ($H(x, p) = \frac{1}{2}|p|^2$), $A_{ij}\partial_{ij}\phi = \Delta\phi$, $f(m)$ nondecreasing and bounded, m_0, m_T smooth and positive
 - Porretta (2013): existence and uniqueness of weak solutions (through energy estimates) for a more general strictly convex Hamiltonian (the Hamiltonian has quadratic growth but no precise asymptotics is required at infinity), $A_{ij}\partial_{ij}\phi = \Delta\phi$, $f(m)$ nondecreasing and bounded, m_0, m_T smooth and positive
 - Achdou, Camilli, Capuzzo Dolcetta (2012) : numerical results
- Idea : Exploit the variational formulation in order to prove existence and uniqueness of weak solutions in the first order case
 - Ok when m_0, m_T are absolutely continuous wrt Lebesgue measure
 - use optimal transport techniques to deal with the case m_0, m_T are general probability Measures

This is an ongoing collaboration with F.Silva