

First order mean-field games

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VIII PARTIAL DIFFERENTIAL EQUATIONS, OPTIMAL DESIGN
AND NUMERICS

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Outline

1 Introduction to Mean Field Games

2 Mean Field Games with state constraints

- The Lagrangian approach
- Existence and uniqueness of relaxed equilibria
- Regularity of relaxed solutions to constrained MFG
- Point-wise properties of relaxed solutions

3 Concluding remarks

- Asymptotic behaviour



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A quote from Wikipedia

*Mean field game theory is the study of strategic decision making in **very large populations** of **small interacting agents**. This class of problems was considered in the **economics** literature by **B Jovanovic** and **RW Rosenthal**, in the **engineering** literature by **PE Caines** and his co-workers, and independently and around the same time by **mathematicians** **J-M Lasry** and **P-L Lions***



Figure: Jovanovic, Rosenthal, Caines, Lasry, and Lions



Motivations for studying MFG

Goal

To describe Nash equilibria in the collective behaviour of a large population of “small” rational agents

- **large population** \rightsquigarrow infinite number (a continuum) of players
- **rational agents** \rightsquigarrow each agent is controlling his/her dynamical own state



Figure: MFG impact: finance, crowd dynamics, smart grids



Lasry-Lions 1: the Hamilton-Jacobi equation

Main idea: to export the principle of statistical mechanics to interactions within rational particles by introducing a **macroscopic description** through a mean field model

- agents are identified with points $x \in \bar{\Omega} \subset \mathbb{R}^n$
- $m(t, dx)$ is the distribution of agents at time t

Agent located in $x \in \bar{\Omega}$ at time $t \in [0, T]$ chooses a path $\gamma_{t,x}(s)$, $s \in [t, T]$, such that

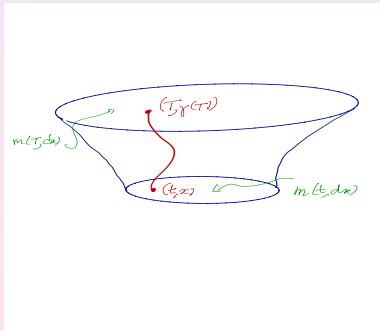
$$u(t, x) := \min_{\gamma(t)=x} \left\{ \int_t^T [L(\gamma(s), \dot{\gamma}(s)) + F(\gamma(s), m(s))] ds + G(\gamma(T), m(T)) \right\}$$

The value function $u(t, x)$ satisfies the associated **Hamilton-Jacobi equation**

$$\begin{cases} -\partial_t u + H(x, \nabla u) = F(x, m) & [0, T] \times \Omega \\ u(T, x) = G(x, m(T)) \end{cases}$$

where

$$H(x, p) := \sup_{v \in \mathbb{R}^n} \{ -\langle p, v \rangle - L(x, v) \}$$



Lasry-Lions 2: the continuity equation

The space gradient $\nabla u(t, x)$ of the solution to the Hamilton-Jacobi equation gives the **optimal feedback** $\gamma_{t,x}$ via the system

$$\gamma'(s) = -\partial_p H(\gamma(s), \nabla u(s, \gamma(s))) \quad (s \in [t, T])$$

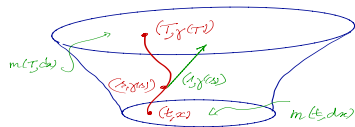
for the minimization problem

$$\min_{\gamma(t)=x} \left\{ \int_t^T [L(\gamma(s), \dot{\gamma}(s)) + F(\gamma(s), m(s))] ds + G(\gamma(T), m(T)) \right\}$$

Since $m(\cdot, dx)$ is just $m_0(dx)$ transported by such a flow, the **continuity equation**

$$\begin{cases} \partial_t m - \operatorname{div}(m \partial_p H(x, \nabla u)) = 0 & [0, T] \times \Omega \\ m(0, dx) = m_0(dx) \end{cases}$$

must be satisfied



Lasry-Lions 3: the MFG system

By coupling the Hamilton-Jacobi equation with the continuity equation above, one obtains the PDE system of **Mean Filed Games**

$$\begin{cases} -\partial_t u + H(x, \nabla u) - F(x, m) = 0 \\ \partial_t m - \operatorname{div}(m \partial_p H(x, \nabla u)) = 0 \end{cases}]0, T[\times \Omega \quad \begin{cases} u(T, x) = G(x, m(T)) \\ m(0, dx) = m_0(dx) \end{cases} \quad (MFG)$$

In the last decade, system (MFG) has been widely investigated for two main kinds of space domains

$$\Omega = \mathbb{T}^n, \mathbb{R}^n$$

main contributions by: Achdou, Bardi, Bensoussan, Camilli, Capuzzo Docetta, Cardaliaguet, Carmona, Delarue, Gomes, Guéant, Lachapelle, Porretta, ...



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Solution of the MFG system

Reference: notes on Mean Field Games
by P. Cardaliaguet, 2013 and 2015

- by vanishing viscosity

$$\begin{cases} -\partial_t u - \epsilon \Delta u + H(x, \nabla u) = F(x, m) \\ \partial_t m - \epsilon \Delta m - \operatorname{div}(m \partial_p H(x, \nabla u_\mu)) = 0 \end{cases}$$



- by a fixed point argument

$$\mu \longrightarrow u_\mu \begin{cases} -\partial_t u + H(x, \nabla u) = F(x, \mu) \\ u(T, x) = G(x, \mu(T)) \end{cases} \longrightarrow m_\mu \begin{cases} \partial_t m - \operatorname{div}(m \partial_p H(x, \nabla u_\mu)) = 0 \\ m(0, dx) = m_0(x) dx \end{cases}$$

Very important facts:

- although nonsmooth, u is linearly semiconcave, which ensures a nice behavior along minimizers
- if m_0 is absolutely continuous with respect to the Lebesgue measure, then $m(\cdot, \cdot)$ stays absolutely continuous



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- although nonsmooth, u is **linearly semiconcave**, which ensures a nice behaviour along minimizers
- if m_0 is absolutely continuous with respect to the Lebesgue measure, then $m(t, \cdot)$ **stays absolutely continuous**



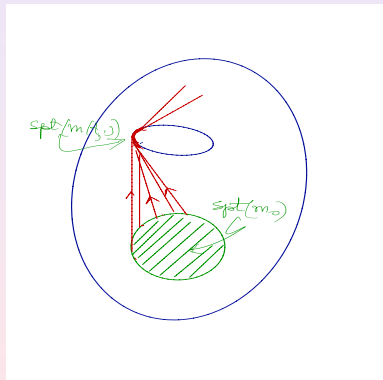
Does all this break down under state constraints?

Our goal To study MFG problems with state constraints: $x \in \bar{\Omega}$

Difficulty

Agent distribution may concentrate
on small sets

Then the above methods break down



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A change of paradigm

- $\Omega \subset \mathbb{R}^n$ bounded domain with boundary of class C^2
- $\mathcal{P}(\bar{\Omega})$ Borel probability measures on $\bar{\Omega}$ with **Katorovich-Rubinstein distance**

$$d_1(m_1, m_2) = \sup \left\{ \int_{\bar{\Omega}} f dm_1 - \int_{\bar{\Omega}} f dm_2 : |f(x) - f(y)| \leq |x - y| \right\}$$

Recall that, given $m \in \mathcal{C}([0, T]; \mathcal{P}(\bar{\Omega}))$, agents aim to attain

$$\min_{\gamma(0)=x, \gamma(t) \in \bar{\Omega}} \left\{ \int_0^T [L(\gamma(t), \dot{\gamma}(t)) + F(\gamma(t), m(t))] dt + G(\gamma(T), m(T)) \right\}$$

but m cannot be fixed a priori as it evolves along optimal feedback

Main idea to overcome such a difficulty:

to replace $m \in \mathcal{C}([0, T]; \mathcal{P}(\bar{\Omega}))$

by a probability measure on the metric space $\mathcal{C}([0, T]; \bar{\Omega})$

that is $\mathcal{C}([0, T]; \mathcal{P}(\bar{\Omega})) \longleftrightarrow \mathcal{P}(\mathcal{C}([0, T]; \bar{\Omega}))$



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Lagrangian approach

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Notation

- constrained arcs

$$\Gamma = \left\{ \gamma \in AC([0, T]; \mathbb{R}^n) : \gamma(t) \in \bar{\Omega}, \forall t \in [0, T] \right\} \quad \text{with } \|\cdot\|_\infty$$

$$\Gamma[x] = \left\{ \gamma \in \Gamma : \gamma(0) = x \right\} \quad (x \in \bar{\Omega})$$

- $\mathcal{P}(\Gamma)$ Borel probability measures on Γ : metric space with d_1 metric

$$d_1(\mu_1, \mu_2) = \sup \left\{ \int_\Gamma f d\mu_1 - \int_\Gamma f d\mu_2 : |f(\gamma) - f(\xi)| \leq \|\gamma - \xi\|_\infty \right\}$$



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Relaxed MFG functional

The evaluation map $e_t : \Gamma \rightarrow \bar{\Omega}$ ($t \in [0, T]$) is defined by $e_t(\gamma) = \gamma(t)$

Push-forward

With any $\mu \in \mathcal{P}(\Gamma)$ and $t \in [0, T]$ one can associate the probability measure $e_t\#\mu$ on $\bar{\Omega}$ given by

$$\int_{\bar{\Omega}} f(x) e_t\#\mu(dx) = \int_{\Gamma} f(\gamma(t)) \mu(d\gamma) \quad \forall f \in C(\bar{\Omega})$$

$e_t\#\mu$ is the push-forward of μ by e_t

For any $\mu \in \mathcal{P}(\Gamma)$ we define

- the associated payoff functional

$$J_\mu[\gamma] = \int_0^T [L(\gamma(t), \dot{\gamma}(t)) + F(\gamma(t), e_t\#\mu)] dt + G(\gamma(T), e_T\#\mu) \quad \forall \gamma \in \Gamma$$

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Relaxed equilibria

A Borel probability measure $\mu \in \mathcal{P}(\Gamma)$ is **compatible with** $m_0 \in \mathcal{P}(\bar{\Omega})$ if

$$e_0 \# \mu = m_0$$

Denote by $\mathcal{P}_{m_0}(\Gamma)$ the subspace consisting of all such measures

Definition

$\mu \in \mathcal{P}_{m_0}(\Gamma)$ is called a **relaxed (CMFG) equilibrium** for m_0 if

$$\text{spt}(\mu) \subseteq \bigcup_{x \in \bar{\Omega}} \Gamma^\mu[x]$$

Equivalently,

$$J_\mu[\bar{\gamma}] = \min_{\gamma \in \Gamma[\bar{\gamma}(0)]} J_\mu[\gamma] \quad \text{for } \mu\text{-a.e. } \bar{\gamma} \in \Gamma$$

where

$$J_\mu[\gamma] = \int_0^T [L(\gamma(t), \dot{\gamma}(t)) + F(\gamma(t), e_t \# \mu)] dt + G(\gamma(T), e_T \# \mu)$$



Relaxed solutions

Let $m_0 \in \mathcal{P}(\bar{\Omega})$

Definition

$(u, m) \in \mathcal{C}([0, T] \times \bar{\Omega}) \times \mathcal{C}([0, T]; \mathcal{P}(\bar{\Omega}))$ is a *relaxed solution* to the CMFG problem if

$$m(t) = e_t \# \mu \quad \forall t \in [0, T]$$

for some relaxed equilibrium $\mu \in \mathcal{P}_{m_0}(\Gamma)$ and

$$u(t, x) = \min_{\gamma \in \Gamma, \gamma(t)=x} \left\{ \int_t^T [L(\gamma(s), \dot{\gamma}(s)) + F(\gamma(s), m(s))] dt + G(\gamma(T), m(T)) \right\}$$



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Existence of relaxed equilibria and solutions

Theorem

For any $m_0 \in \mathcal{P}(\bar{\Omega})$ there is at least one relaxed equilibrium

Corollary

For any $m_0 \in \mathcal{P}(\bar{\Omega})$ there is at least one relaxed solution (u, m) to the CMFG problem



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Proof

Kakutani's fixed-point theorem

- $S \neq \emptyset$ compact convex subset of a locally convex Hausdorff space
 - $\phi : S \rightrightarrows S$ nonempty convex-valued with closed graph
- $\implies \phi$ has a fixed point.

Proof of theorem: construction of a fixed point of $E : \mathcal{P}_{m_0}(\Gamma) \rightrightarrows \mathcal{P}_{m_0}(\Gamma)$

$$E(\eta) = \left\{ \mu \in \mathcal{P}_{m_0}(\Gamma) \mid \text{spt}(\mu_x) \subseteq \Gamma^\eta[x] \text{ for } m_0\text{-a.e. } x \in \bar{\Omega} \right\} \quad (\eta \in \mathcal{P}_{m_0}(\Gamma))$$

where $\{\mu_x\}_{x \in \bar{\Omega}} \subset \mathcal{P}(\Gamma)$ is the family of probability measures which disintegrates μ

$$\mu = \int_{\bar{\Omega}} \mu_x dm_0(x) \quad \text{and} \quad \text{spt}(\mu_x) \subseteq \Gamma[x] \quad m_0\text{-a.e. } x \in \bar{\Omega}$$

Indeed

$$\eta \in \mathcal{P}_{m_0}(\Gamma) \text{ relaxed equilibrium} \iff \eta \in E(\eta)$$

The existence of a fixed point of E follows from Kakutani's Theorem

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$$E(\eta) = \left\{ \mu \in \mathcal{P}_{m_0}(\Gamma) \mid \text{spt}(\mu_x) \subseteq \Gamma^\eta[x] \text{ for } m_0\text{-a.e. } x \in \bar{\Omega} \right\} \quad (\eta \in \mathcal{P}_{m_0}(\Gamma))$$

where $\{\mu_x\}_{x \in \bar{\Omega}} \subset \mathcal{P}(\Gamma)$ is the family of probability measures which **disintegrates** μ

$$\mu = \int_{\bar{\Omega}} \mu_x dm_0(x) \quad \text{and} \quad \text{spt}(\mu_x) \subseteq \Gamma[x] \quad m_0\text{-a.e. } x \in \bar{\Omega}$$

Indeed

$$\eta \in \mathcal{P}_{m_0}(\Gamma) \text{ relaxed equilibrium} \iff \eta \in E(\eta)$$

The existence of a fixed point of E follows from **Kakutani's Theorem**



Uniqueness

Theorem

Assume *monotonicity conditions*: for any $m_1, m_2 \in \mathcal{P}(\bar{\Omega})$

$$\begin{cases} \int_{\bar{\Omega}} (G(x, m_1) - G(x, m_2)) d(m_1 - m_2)(x) \geq 0 \\ \int_{\bar{\Omega}} (F(x, m_1) - F(x, m_2)) d(m_1 - m_2)(x) > 0 \quad \text{if } m_1 \neq m_2 \end{cases}$$

If (u_1, m_1) and (u_2, m_2) are relaxed solutions to the CMFG problem, then

$$u_1 \equiv u_2 \quad \text{and} \quad m_1 = m_2$$

F satisfies the strict monotonicity condition if $F : \bar{\Omega} \times \mathcal{P}(\bar{\Omega}) \rightarrow \mathbb{R}$ is of the form

$$F(x, m) = \int_{\bar{\Omega}} f(y, (\phi \star m)(y)) \phi(x - y) dy$$

where $\phi : \mathbb{R}^d \rightarrow \mathbb{R}$ is a smooth even kernel with compact support and

$f : \bar{\Omega} \times \mathbb{R} \rightarrow \mathbb{R}$ is smooth and $f(x, \cdot)$ is strictly increasing



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Outline

- 1 Introduction to Mean Field Games
- 2 Mean Field Games with state constraints
 - The Lagrangian approach
 - Existence and uniqueness of relaxed equilibria
 - **Regularity of relaxed solutions to constrained MFG**
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More notation and assumptions

Recall $\Omega \subset \mathbb{R}^n$ is bounded with $\partial\Omega \in \mathcal{C}^2$. Consequently

- **distance**

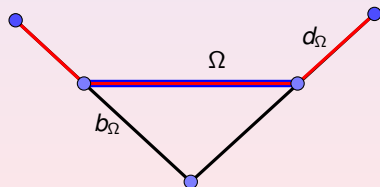
$$d_\Omega(x) = \min_{y \in \bar{\Omega}} |x - y|$$

of class $\mathcal{C}^2(\Omega_\delta^+)$ for some $\delta > 0$ with $\Omega_\delta^+ = \{x \in \mathbb{R}^n \setminus \Omega : d_\Omega(x) < \delta\}$

- **oriented boundary distance**

$$b_\Omega(x) = d_\Omega(x) - d_{\mathbb{R}^n \setminus \Omega}(x)$$

of class $\mathcal{C}^2(\Omega_\delta)$ on $\Omega_\delta = \{x \in \mathbb{R}^n : |b_\Omega(x)| < \delta\}$



References

- Dubovitskii – Milyutin (1964)
- Malanowski (1978)
- Hager (1979)
- Vinter (2000)
- Galbraith – Vinter (2003)
- Frankowska (2006, 2009)
- Bettiol – Frankowska (2007, 2008)
- Bettiol – Khalil – Vinter (2016)



Necessary conditions for smooth state constraints

Theorem

Given $x \in \bar{\Omega}$ let γ^* minimize over $\Gamma[x]$ the functional

$$\gamma \mapsto \int_0^T [L(\gamma(s), \dot{\gamma}(s)) + f(s, \gamma(s))] dt + g(\gamma(T))$$

where $g \in C^1(\bar{\Omega})$ and $f : [0, T] \times \bar{\Omega} \rightarrow \mathbb{R}$ satisfies $|f_t| + |\nabla f| \leq C$

Then there exist

- $p^* : [0, T] \rightarrow \mathbb{R}^n$ Lipschitz
- $\nu \in \mathbb{R}$ and $\Lambda \in C_b([0, T] \times \Omega_\delta \times \mathbb{R}^n)$ (independent of γ^*, p^*)

such that ($\mathbb{I}_{\partial\Omega}$ = characteristic function of $\partial\Omega$)

$$\begin{cases} \dot{\gamma}^* = -\partial_p H(\gamma^*, p^*) \\ \dot{p}^* = \nabla H(\gamma^*, p^*) - \nabla f(t, \gamma^*) - \Lambda(t, \gamma^*, p^*) \mathbb{I}_{\partial\Omega}(\gamma^*) \nabla b_\Omega(\gamma^*) \\ p^*(T) = \nabla g(\gamma^*(T)) + \nu \mathbb{I}_{\partial\Omega}(\gamma^*(T)) \nabla b_\Omega(\gamma^*(T)) \end{cases} \quad \forall t \in [0, T]$$

Consequently, $\gamma^* \in C_{Lip}^1([0, T]; \mathbb{R}^n)$ and $\|\dot{\gamma}^*\|_{Lip} \leq C(\Omega, H, f, g)$

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Existence of Lipschitz solutions

Theorem

Let $m_0 \in \mathcal{P}(\bar{\Omega})$ and suppose

$$|F(x_1, m_1) - F(x_2, m_2)| + |G(x_1, m_1) - G(x_2, m_2)| \leq C(|x_1 - x_2| + d_1(m_1, m_2))$$

Then there exists at least one relaxed solution of CMFG problem (u, m) such that

$$u \in Lip([0, T] \times \bar{\Omega}) \quad \text{and} \quad m \in Lip([0, T]; \mathcal{P}(\bar{\Omega}))$$

Such a solution will be called a *Lipschitz relaxed solution* of the CMFG problem

The proof applies necessary conditions to construct a relaxed CMFG equilibrium

$$\eta \in \mathcal{P}_{m_0}(\Gamma) \text{ such that } m(t) := e_t \# \eta \text{ belongs to } Lip([0, T]; \mathcal{P}(\bar{\Omega}))$$

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A quick look at semiconcave functions

$\Omega \subseteq \mathbb{R}^n$ open

$v : \Omega \rightarrow \mathbb{R}$ **semiconcave with modulus** $\omega : [0, \infty[\rightarrow [0, \infty[$ if

$$\lambda v(x) + (1 - \lambda)v(y) - v(\lambda x + (1 - \lambda)y) \leq \lambda(1 - \lambda)|x - y|\omega(|x - y|)$$

for all x, y such that $[x, y] \subset \Omega$ and $\lambda \in [0, 1]$

Special cases:

- $\omega(s) \equiv 0 \rightarrow$ **concave**
- $\omega(s) = Cs$ ($C > 0$) \rightarrow **linearly semiconcave**

In this case

$$x \mapsto v(x) - \frac{C}{2}|x|^2 \tag{*}$$

is concave on all convex subsets of Ω

- $\omega(s) = Cs^\alpha$ ($C > 0, 0 < \alpha < 1$) \rightarrow **fractionally semiconcave**

In this case, (*) is no longer valid



Some references on semiconcave functions

- **control theory and sensitivity analysis**
Hrustalev 1978, C – Frankowska 1991
Fleming – McEneaney 2000
Rifford 2000, 2002
- **nonsmooth and variational analysis**
Rockafellar 1982
Colombo – Marigonda 2006, Colombo – Nguyen 2010
- **differential geometry**
Perelman 1995, Petrunin 2007
- **monographs**
C – Sinestrari (Birkhäuser 2004)
Villani (Springer 2009)



Semiconcavity & nonsmooth analysis

For any semiconcave $v : \Omega \rightarrow \mathbb{R}$

- the **superdifferential** at $x \in \Omega$ coincides with Clarke's gradient

$$D^+ v(x) = \text{co } D^* v(x) = \partial v(x)$$

where $D^* v(x) = \{ \lim_{i \rightarrow \infty} Dv(x_i) \mid x_i \rightarrow x \}$ **reachable gradients**

- $D^+ v(x) = \{p\} \iff v$ **differentiable**



Semiconcavity of relaxed Lipschitz solution

Theorem

Any Lipschitz relaxed solution (u, m) of CMFG problem is *locally semiconcave* on $[0, T[\times \bar{\Omega}$ with a *fractional modulus*:

$\forall \rho \in]0, T[$ there exists $C_\rho \geq 0$ such that

$$u(t + \tau, x + h) + u(t - \tau, x - h) - 2u(t, x) \leq C_\rho (|\tau| + |h|)^{3/2}$$

for all $t, t \pm \tau \in [0, T - \rho]$ and $x, x \pm h \in \bar{\Omega}$

Several proofs of the above result can be given

An interesting method of proof uses *sensitivity relations* that we discuss next



Adjoint state inclusion / sensitivity relations

Given

- a Lipschitz relaxed solution (u, m) of the CMFG problem
- $(t, x) \in [0, T[\times \bar{\Omega}$ and a solution $\gamma^* \in \Gamma$ to

$$\min_{\gamma \in \Gamma, \gamma(t) = x} \left\{ \int_t^T [L(\gamma(s), \dot{\gamma}(s)) + F(\gamma(s), m(s))] dt + G(\gamma(T), m(T)) \right\}$$

- the adjoint state $p^* : [t, T] \rightarrow \mathbb{R}^n$ associated with γ^*

we have that

$$\left(H(\gamma^*(s), p^*(s)) - F(\gamma^*(s), m(s)), p^*(s) \right) \in D^+ u(s, \gamma^*(s)) \quad \forall s \in [t, T[$$

and $\forall \rho \in]0, T[$ there exists $C_\rho \geq 0$ such that $\forall t, t + \tau \in [0, T - \rho]$ and all $x + h \in \bar{\Omega}$

$$\begin{aligned} u(t + \tau, x + h) - u(t, x) - \tau(H(x, p^*(t)) - F(x, m(t))) - \langle p^*(t), h \rangle \\ \leq C_\rho (|\tau| + |h|)^{3/2} \end{aligned}$$



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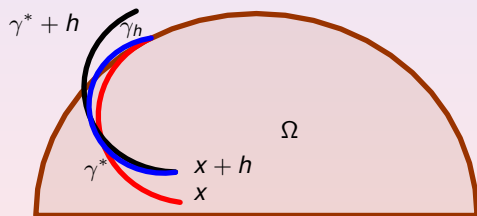
Proof of sensitivity relation for $\tau = 0$

We want to show that $\forall t \in [0, T - \rho]$ and all $x + h \in \bar{\Omega}$

$$u(t, x + h) - u(t, x) - \langle p(t), h \rangle \leq C_\rho |h|^{3/2}$$

Let $0 < \sigma \leq \rho$ to be fixed later and define for all $s \in [t, T]$

$$\gamma_h(s) = \gamma^*(s) + \left(1 + \frac{t-s}{\sigma}\right)_+ h$$



$$\hat{\gamma}_h(s) = \gamma_h(s) - d_{\bar{\Omega}}(\gamma_h(s)) Dd_{\partial\Omega}(\gamma_h(s))$$



Proof of sensitivity relation (continued)

By dynamic programming

$$\begin{aligned}
 u(t, x + h) - u(t, x) - \langle p(t), h \rangle &\leq \int_t^{t+\sigma} [L(\hat{\gamma}_h, \dot{\hat{\gamma}}_h) - L(\gamma^*, \dot{\gamma}^*)] ds \\
 &\quad + \int_t^{t+\sigma} [F(\hat{\gamma}_h, m) - F(\gamma^*, m)] ds - \langle p(t), h \rangle
 \end{aligned} \tag{1}$$

We want to express $\langle p(t), h \rangle$ so we expand

$$\begin{aligned}
 -\langle p(t), h \rangle &= -\langle p(t + \sigma), \underbrace{\hat{\gamma}_h(t + \sigma) - \gamma^*(t + \sigma)}_{=0} \rangle + \int_t^{t+\sigma} \frac{d}{ds} \langle p, \hat{\gamma}_h - \gamma^* \rangle ds \\
 &= \int_t^{t+\sigma} \langle \dot{p}, \hat{\gamma}_h - \gamma^* \rangle ds + \int_t^{t+\sigma} \langle p, \dot{\hat{\gamma}}_h - \dot{\gamma}^* \rangle ds
 \end{aligned}$$

By appealing to PMP to represent $\langle \dot{p}, \hat{\gamma}_h - \gamma^* \rangle$ and $\langle p, \dot{\hat{\gamma}}_h - \dot{\gamma}^* \rangle$ we obtain

$$\begin{aligned}
 u(t, x + h) - u(t, x) - \langle p(t), h \rangle &\leq \dots \\
 &\leq C \int_t^{t+\sigma} |\hat{\gamma}_h - \gamma^*|^2 ds + C \int_t^{t+\sigma} |\dot{\hat{\gamma}}_h - \dot{\gamma}^*|^2 ds + C \int_t^{t+\sigma} |\hat{\gamma}_h - \gamma^*| ds
 \end{aligned}$$



Proof of sensitivity relation (completed)

Recalling

$$\begin{cases} \gamma_h(s) = \gamma^*(s) + \left(1 + \frac{t-s}{\sigma}\right)_+ h \\ \hat{\gamma}_h(s) = \gamma_h(s) - d_{\bar{\Omega}}(\gamma_h(s)) Dd_{\partial\Omega}(\gamma_h(s)) \end{cases}$$

we have that

$$|\hat{\gamma}_h(s) - \gamma^*(s)| \leq 2|h| \quad \forall s \in [t, t + \sigma]$$

Using the regularity of the distance functions one can also prove (technical)

$$\int_t^{t+\sigma} |\dot{\hat{\gamma}}_h(s) - \dot{\gamma}^*(s)|^2 ds \leq C \frac{|h|^2}{\sigma} + C|h|\sigma$$

Therefore

$$u(t, x + h) - u(t, x) - \langle p(t), h \rangle \leq C|h| \left(\frac{|h|}{\sigma} + \sigma \right) \leq 2C|h|^{3/2}$$

by taking $\sigma = |h|^{1/2}$



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Point-wise solutions of the HJ equation

Given a Lipschitz relaxed solution (u, m) to CMFG problem, we have that

(I) u is a **constrained viscosity solution** of

$$\begin{cases} -\partial_t u + H(x, \nabla u) = F(x, m) & \text{in }]0, T[\times \bar{\Omega} \\ u(T, x) = G(x, m(T)) & \forall x \in \bar{\Omega} \end{cases}$$

Moreover, defining

$$\begin{aligned} Q_m &= \left\{ (t, x) \in]0, T[\times \Omega : x \in \text{spt}(m(t)) \right\} \\ \partial Q_m &= \left\{ (t, x) \in]0, T[\times \partial\Omega : x \in \text{spt}(m(t)) \right\} \end{aligned}$$

the following holds true

(II) u is differentiable on Q_m and $-\partial_t u + H(x, \nabla u) = F(x, m)$ on Q_m

(III) u has

time derivative, one-sided normal derivative, and tangential gradient on ∂Q_m

(IV) the tangential gradient $\nabla^\tau u$ satisfies

$$-\partial_t u + H^\tau(x, \nabla_x^\tau u) = F(x, m) \text{ on } \partial Q_m$$

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(II) u is **differentiable** on Q_m and $-\partial_t u + H(x, \nabla u) = F(x, m)$ on Q_m

(III) u has

time derivative, one-sided **normal derivative**, and **tangential gradient** on ∂Q_m

(IV) the tangential gradient $\nabla^\tau u$ satisfies

$$-\partial_t u + H^\tau(x, \nabla_x^\tau u) = F(x, m) \text{ on } \partial Q_m$$

where $H^\tau(x, p) = \sup \left\{ -\langle p, v \rangle - L(x, v) \mid \langle v, \nu(x) \rangle = 0 \right\}$



Analysis of the continuity equation

Given a Lipschitz relaxed solution (u, m) to CMFG problem, we have that

- (I) there exists a **bounded continuous** vector field $V :]0, T] \times \bar{\Omega} \rightarrow \mathbb{R}^n$ such that m satisfies the continuity equation

$$\partial_t m + \operatorname{div}(mV) = 0 \quad \text{in }]0, T[\times \bar{\Omega}$$

in the sense of distributions: $\forall \phi \in C_c^1(]0, T[\times \bar{\Omega})$

$$\int_0^T \int_{\bar{\Omega}} (\phi_t + \langle V, \nabla \phi \rangle) dm(t, dx) dt = 0$$

- (II) V is given by the **optimal feedback** on Q_m , that is,

$$V(t, x) = \begin{cases} -\partial_p H(x, \nabla u(t, x)) & \forall (t, x) \in Q_m \\ -\partial_p H(x, \nabla_x^\tau u(t, x) + \partial_{v_i}^+ u(t, x) \nu_i(x)) & \forall (t, x) \in \partial Q_m \end{cases}$$



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Proof

- Consider the continuous map $V_m : Q_m \cup \partial Q_m \rightarrow \mathbb{R}^n$

$$V_m(t, x) = \begin{cases} -\partial_p H(x, \nabla u(t, x)) & \forall (t, x) \in Q_m \\ -\partial_p H(x, \nabla_x^\tau u(t, x) + \partial_{\nu_i}^+ u(t, x) \nu_i(x)) & \forall (t, x) \in \partial Q_m \end{cases}$$

and extend it to a continuous vector field $V :]0, T[\times \bar{\Omega} \rightarrow \mathbb{R}^n$ by **Tietze theorem**

- Let η be a **constrained equilibrium** associated with (u, m) : then

$$(t, \gamma(t)) \in Q_m \cup \partial Q_m \quad \text{and} \quad \dot{\gamma}(t) = V(t, \gamma(t)) \quad \forall t \in]0, T[$$

for η -a.e. $\gamma \in \Gamma$

- So, $\forall \phi \in \mathcal{C}_c^1(]0, T[\times \bar{\Omega})$ we use the change of variables $m(t) = e_{t\#} \eta$ to compute

$$\begin{aligned} \frac{d}{dt} \int_{\bar{\Omega}} \phi(t, x) m(t, dx) &= \frac{d}{dt} \int_{\Gamma} \phi(t, \gamma(t)) \eta(d\gamma) \\ &= \int_{\Gamma} (\partial_t \phi(t, \gamma(t)) + \langle D\phi(t, \gamma(t)), \underbrace{\dot{\gamma}(t)}_{=V(t, \gamma(t))} \rangle) \eta(d\gamma) \\ &= \int_{\bar{\Omega}} (\partial_t \phi(t, x) + \langle D\phi(t, x), V(t, x) \rangle) m(t, dx) \end{aligned}$$



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Conclusions

We have shown how to recover a fairly complete theory for the

- existence and uniqueness
- regularity
- pointwise behaviour

of solutions to constrained MFG systems

This opens the way to the study of at least two main problems

- Since constrained equilibria may develop singular parts (Dirac masses) induced by the presence of state constraints, are such singularities stable or do they disappear if constraints become inactive?
- How to describe the behaviour of the solution (u^T, m^T) of the constrained Mean Field Games system

$$\begin{cases} -\partial_t u^T(t, x) + H(x, \nabla_x u^T(t, x)) = F(x, m^T(t)), & \text{in }]0, T[\times \bar{\Omega} \\ \partial_t m^T(t) - \operatorname{div}(m^T(t) D_p H(x, \nabla_x u^T(t, x))) = 0, & \text{in }]0, T[\times \bar{\Omega} \\ u^T(T, x) = u^f(x), \quad m^T(0) = m_0, & \text{in } \bar{\Omega}. \end{cases} \quad (\text{CMFG})$$

as $T \rightarrow +\infty$?



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Outline

- 1 Introduction to Mean Field Games
- 2 Mean Field Games with state constraints
 - The Lagrangian approach
 - Existence and uniqueness of relaxed equilibria
 - Regularity of relaxed solutions to constrained MFG
 - Point-wise properties of relaxed solutions
- 3 Concluding remarks
 - Asymptotic behaviour



Asymptotic behaviour: the unconstrained case

References

- (i) P. Cardaliaguet (2013) on \mathbb{T}^n
- (ii) joint work with W. Cheng, C. Mendico, and K. Wang (2019) in \mathbb{R}^n
under the following assumptions

(F1) There is a constant $C > 0$ such that for every $m_1, m_2 \in \mathcal{P}_1(\mathbb{R}^n)$,

$$\int_{\mathbb{R}^n} (F(x, m_1) - F(x, m_2)) d(m_1 - m_2) \geq C \int_{\mathbb{R}^n} (F(x, m_1) - F(x, m_2))^2 dx$$

(F2) There exist a compact set $K_0 \subset \mathbb{R}^n$ and a constant $\delta_0 > 0$ such that

$$\min_{x \in K_0} \left\{ L(x, 0) + F(x, m) \right\} \leq \inf_{x \in \mathbb{R}^n \setminus K_0} \left\{ L(x, 0) + F(x, m) \right\} - \delta_0, \quad \forall m \in \mathcal{P}_1(\mathbb{R}^n)$$



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Ergodic Mean Field Games system

The limit behaviour of solutions is captured by

Ergodic Mean Field Games (EMFG) system

$$\begin{cases} H(x, D\bar{u}(x)) = c_H(\bar{m}) + F(x, \bar{m}) & \text{in } \mathbb{R}^n \\ \operatorname{div}(\bar{m} \nabla_p H(x, D\bar{u}(x))) = 0 & \text{in } \mathbb{R}^n \\ \int_{\mathbb{R}^n} \bar{m}(dx) = 1 \end{cases}$$

where Mañé's critical value $c_H(\bar{m})$ is defined by

$$c_H(\bar{m}) := \inf \{c \in \mathbb{R} : \exists u \in C(\mathbb{R}^n) \text{ viscosity solution of } H(x, Du) = c + F(x, \bar{m})\}$$

see A. Fathi, "Weak KAM Theorem in Lagrangian dynamics"



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Solution of (*EMFG*)

$$\begin{cases} H(x, D\bar{u}(x)) = c_H(\bar{m}) + F(x, \bar{m}), & \text{in } \mathbb{R}^n \\ \operatorname{div}(\bar{m} \nabla_p H(x, D\bar{u}(x))) = 0, & \text{in } \mathbb{R}^n \\ \int_{\mathbb{R}^n} \bar{m}(dx) = 1. \end{cases} \quad (\text{EMFG})$$

Theorem (existence of solutions – uniqueness of critical values)

- (i) *There exists at least one solution $(\bar{u}, \bar{m}, c_H(\bar{m}))$ of system EMFG*
- (ii) *Let $(\bar{u}_1, \bar{m}_1, c_H(\bar{m}_1)), (\bar{u}_2, \bar{m}_2, c_H(\bar{m}_2))$ solve (EMFG). Then,*

$$c_H(\bar{m}_1) = c_H(\bar{m}_2) \quad \text{and} \quad F(x, \bar{m}_1) = F(x, \bar{m}_2), \quad \forall x \in \mathbb{R}^n$$



Convergence of MFG solution

Theorem

Let $(\bar{u}, \bar{m}, c_H(\bar{m}))$ be any solution of

$$\begin{cases} H(x, D\bar{u}(x)) = c_H(\bar{m}) + F(x, \bar{m}), & \text{in } \mathbb{R}^n \\ \operatorname{div}(\bar{m} \nabla_\rho H(x, D\bar{u}(x))) = 0, & \text{in } \mathbb{R}^n \\ \int_{\mathbb{R}^n} \bar{m}(dx) = 1. \end{cases} \quad (EMFG)$$

Then, for any sufficiently large $R > 0$ there exists a constant $C(R) > 0$ such that for every $T \geq 1$ the solution (u^T, m^T) of the MFG system satisfies

$$\sup_{t \in [0, T]} \frac{\|u^T(t, \cdot) - c_H(\bar{m})(t - T)\|_{\infty, \bar{B}_R}}{T} \leq \frac{C(R)}{T^{\frac{1}{n+2}}}, \quad (2)$$

$$\frac{1}{T} \int_0^T \|F(\cdot, m^T(s)) - F(\cdot, \bar{m})\|_{\infty, \bar{B}_R} ds \leq \frac{C(R)}{T^{\frac{1}{n+2}}}. \quad (3)$$

Thank you for your attention!



Figure: Rational agents at work, Benasque 2019

