

Maximum principle for optimal control problem of non exchangeable mean field systems

Samy Mekkaoui

CMAP, École Polytechnique and LPSM

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Joint work with

Idris Kharroubi (LPSM, Sorbonne Université)

Huỳnh Pham (CMAP, École Polytechnique)

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2 Some preliminaries tools on $L^2(\mathcal{P}_2(\mathbb{R}^d))$

3 Pontryagin principle for optimality

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5 An application to linear quadratic mean field control

Classical MFC problem

Classical mean-field control (MFC) problem :

$$\inf_{\alpha \in \mathcal{A}} J(\alpha) := \mathbb{E} \left[\int_0^T f(X_t^\alpha, \mathbb{P}_{X_t^\alpha}, \alpha_t) dt + g(X_T, \mathbb{P}_{X_T^\alpha}) \right], \quad (1)$$

where \mathcal{A} defines a suitable class of control with controlled state $X^\alpha = (X_t^\alpha)_{t \in [0, T]}$ dynamics given by

$$\begin{aligned} dX_t^\alpha &= b(X_t^\alpha, \mathbb{P}_{X_t^\alpha}, \alpha_t) dt + \sigma(X_t^\alpha, \mathbb{P}_{X_t^\alpha}, \alpha_t) dW_t, \\ X_0^\alpha &= \xi, \end{aligned} \quad (2)$$

where the random variables are defined on a filtered probability space $(\Omega, \mathcal{F}, \mathbb{F}, \mathbb{P})$ supporting a brownian motion W and an initial random variable ξ .

→ 2 well known methods to study (1)-(2) : DPP and Pontryagin maximum principle.

Introduction

Context and motivations

→ Extend the known *MFG/MFC* theory to non exchangeable interactions. A lot of literature has been developed recently within the graphons theory (see the works of Bayraktar et al. and Aurell et al. for instance) where an agent labeled by $u \in I := [0, 1]$ interacts with the other agents through the probability measure

$$\frac{\int_I G(u, v) \mathbb{P}_{x_t^v}(\mathrm{d}x) \mathrm{d}v}{\int_I G(u, v) \mathrm{d}v} \in \mathcal{P}(\mathbb{R}^d), \quad 0 \leq t \leq T, \quad u \in I. \quad (3)$$

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→ Extend the framework without specifying the type of interaction. Dynamics are functions of the collection of laws $(\mathbb{P}_{X_t^v})_{v \in I}$.
(see De Crescenzo, Fuhrman, Kharroubi and Pham [2] for the first introduction to this framework).

Introduction

The NE-MFC problem

- Central planner aims to control a system of interacting heterogeneous agents :

Non Exchangeable Mean Field SDE

$$\begin{aligned} dX_t^u &= b(u, X_t^u, \alpha_t^u, (\mathbb{P}_{X_t^v})_{v \in I}) dt + \sigma(u, X_t^u, \alpha_t^u, (\mathbb{P}_{X_t^v})_{v \in I}) dW_t^u, \quad 0 \leq t \leq T, u \in I, \quad (4) \\ X_0^u &= \xi^u. \end{aligned}$$

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$$X_0^u = \xi^u.$$

→ Minimize over a collection of processes $\alpha = (\alpha^u)_{u \in I}$ in a suitable class \mathcal{A} the following cost functional :

Cost Functional

$$J(\alpha) = \int_I \mathbb{E} \left[\int_0^T f(u, X_t^u, \alpha_t^u, (\mathbb{P}_{X_t^v})_{v \in I})dt + g(u, X_T^u, (\mathbb{P}_{X_T^v})_{v \in I}) \right] du \quad (5)$$

→ Compute $V_0 = J(\alpha^*)$ where α^* is a minimizer of J .

Introduction

Goal of this presentation

Objectives :

- Adapt the **Pontryagin Maximum Principle** to mean field control for non exchangeable mean field systems (NE-MFC) to find necessary and sufficient conditions for an admissible optimal control α .

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- Adapt the **Pontryagin Maximum Principle** to mean field control for non exchangeable mean field systems (NE-MFC) to find necessary and sufficient conditions for an admissible optimal control α .
- Propose an illustration in the **Linear Quadratic (LQ)** case.

We notice that independantly of our work, Cao and Laurière in [6] study also the **Pontryagin Maximum Principle** in the context of nonlinear graphon-based interactions, i.e., the coefficients b , σ depend on the collection of laws only through a graphon-weighted measure.

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- 2 Some preliminaries tools on $L^2(\mathcal{P}_2(\mathbb{R}^d))$
- 3 Pontryagin principle for optimality
- 4 Solvability of non exchangeable mean field FBSDEs
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Analysis tools on $L^2(\mathcal{P}_2(\mathbb{R}^d))$

Definition of $L^2(\mathcal{P}_2(\mathbb{R}^d))$

Following the *NE – MFC* setting, we need to introduce a suitable space for the collection of measures which we will denote as $L^2(I; \mathcal{P}_2(\mathbb{R}^d)) := L^2(\mathcal{P}_2(\mathbb{R}^d))$.

Definition of $L^2(\mathcal{P}_2(\mathbb{R}^d))$

The space $L^2(\mathcal{P}_2(\mathbb{R}^d))$ is defined as

$$\left\{ \mu = (\mu^u)_{u \in I} \text{ s.t. } I \ni u \mapsto \mu^u \in \mathcal{P}_2(\mathbb{R}^d) \text{ is measurable and } \int_I \int_{\mathbb{R}^d} |x|^2 \mu^u(dx) du < \infty \right\}.$$

- The space $L^2(\mathcal{P}_2(\mathbb{R}^d))$ is endowed with the metric :

$$\mathbf{d}(\mu, \nu) := \int_I \mathcal{W}_2(\mu^u, \nu^u)^2 du, \quad \mu := (\mu^u)_{u \in I}, \quad \nu := (\nu^u)_{u \in I}. \quad (6)$$

- Functions b and σ are now defined on the set $L^2(\mathcal{P}_2(\mathbb{R}^d))$

Analysis tools on $L^2(\mathcal{P}_2(\mathbb{R}^d))$

A notion of derivative

A derivative in $L^2(\mathcal{P}_2(\mathbb{R}^d))$ (1)

(i) Given a function $v : L^2(\mathcal{P}_2(\mathbb{R}^d)) \rightarrow \mathbb{R}$, we say that a measurable function

$$\frac{\delta}{\delta m} v : L^2(\mathcal{P}_2(\mathbb{R}^d)) \times I \times \mathbb{R}^d \ni (\mu, u, x) \mapsto \frac{\delta}{\delta m} v(\mu)(u, x) \in \mathbb{R} \quad (7)$$

is the linear functional derivative (or flat derivative) of v if

1. $(\mu, x) \mapsto \frac{\delta}{\delta m} v(\mu)(u, x)$ is continuous from $L^2(\mathcal{P}_2(\mathbb{R}^d)) \times \mathbb{R}^d$ to \mathbb{R} for all $u \in I$;
2. for every compact set $K \subset L^2(\mathcal{P}_2(\mathbb{R}^d))$ there exists a constant $C_K > 0$ such that

$$\left| \frac{\delta}{\delta m} v(\mu)(u, x) \right| \leq C_K (1 + |x|^2),$$

for all $u \in I$, $x \in \mathbb{R}^d$, $\mu \in K$;

3. we have

$$\begin{aligned} v(\nu) - v(\mu) &= \int_0^1 \left\langle \frac{\delta}{\delta m} v(\mu + \theta(\nu - \mu)), \nu - \mu \right\rangle d\theta \\ &= \int_0^1 \int_I \int_{\mathbb{R}^d} \frac{\delta}{\delta m} v(\mu + \theta(\nu - \mu))(u, x) (\nu^u - \mu^u)(dx) du d\theta \end{aligned}$$

for all $\mu, \nu \in L^2(\mathcal{P}_2(\mathbb{R}^d))$.

Analysis tools on $L^2(\mathcal{P}_2(\mathbb{R}^d))$

A notion of derivative

A derivative in $L^2(\mathcal{P}_2(\mathbb{R}^d))$ (2)

(ii) We say that the function v admits a continuously differentiable flat derivative if

1. v admits a flat derivative $\frac{\delta}{\delta m} v$ satisfying $x \mapsto \frac{\delta}{\delta m} v(\mu)(u, x)$ is Fréchet differentiable with Fréchet derivative denoted by $x \mapsto \partial \frac{\delta}{\delta m} v(\mu)(u, x)$ for all $(\mu, u) \in L^2(\mathcal{P}_2(\mathbb{R}^d)) \times I$;
2. $(\mu, x) \mapsto \partial \frac{\delta}{\delta m} v(\mu)(u, x)$ is continuous from $L^2(\mathcal{P}_2(\mathbb{R}^d)) \times \mathbb{R}^d$ to \mathbb{R} for all $u \in I$;
3. for every compact set $K \subset L^2(\mathcal{P}_2(\mathbb{R}^d))$ there exists a constant $C_K > 0$ such that

$$\left| \partial \frac{\delta}{\delta m} v(\mu)(u, x) \right| \leq C_K (1 + |x|^2),$$

for all $u \in I$, $x \in \mathbb{R}^d$, $\mu \in K$.

Analysis tools on $L^2(\mathcal{P}_2(\mathbb{R}^d))$

Gâteaux derivatives

Gâteaux derivative on $L^2(\mathcal{P}_2(\mathbb{R}^d))$

Let $f : I \times \mathbb{R}^d \times L^2(\mathcal{P}_2(\mathbb{R}^d)) \rightarrow \mathbb{R}$ assumed to have a continuously differentiable linear functional derivative $\partial \frac{\delta}{\delta m} f$. For $X, Y \in L^2(\Omega, \mathcal{F}, \mathbb{P}, \mathbb{R}^d)'$ such that $(\mathbb{P}_{X^v})_{v \in I}, (\mathbb{P}_{Y^v})_{v \in I} \in L^2(\mathcal{P}_2(\mathbb{R}^d))$ we have

$$\lim_{\epsilon \rightarrow 0} \frac{1}{\epsilon} (f(\mathbf{u}, \mathbf{x}, (\mathbb{P}_{X^v + \epsilon Y^v})_{v \in I}) - f(\mathbf{u}, \mathbf{x}, (\mathbb{P}_{X^v})_{v \in I})) = \int_I \mathbb{E} \left[\partial \frac{\delta}{\delta m} f(\mathbf{u}, \mathbf{x}, (\mathbb{P}_{X^v})_{v \in I})(\tilde{u}, X^{\tilde{u}}) \cdot Y^{\tilde{u}} \right] d\tilde{u} \quad (8)$$

Analysis tools on $L^2(\mathcal{P}_2(\mathbb{R}^d))$

Gateaux derivatives

Gateaux derivative on $L^2(\mathcal{P}_2(\mathbb{R}^d))$

Let $f : I \times \mathbb{R}^d \times L^2(\mathcal{P}_2(\mathbb{R}^d)) \rightarrow \mathbb{R}$ assumed to have a continuously differentiable linear functional derivative $\partial \frac{\delta}{\delta m} f$. For $X, Y \in L^2(\Omega, \mathcal{F}, \mathbb{P}, \mathbb{R}^d)'$ such that $(\mathbb{P}_{X^v})_{v \in I}, (\mathbb{P}_{Y^v})_{v \in I} \in L^2(\mathcal{P}_2(\mathbb{R}^d))$ we have

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→ Relation (8) is understood as a calculus of variation on $L^2(\mathcal{P}_2(\mathbb{R}^d))$.

Analysis tools on $L^2(\mathcal{P}_2(\mathbb{R}^d))$

A notion of convexity

A notion of convexity in $L^2(\mathcal{P}_2(\mathbb{R}^d))$

Let $f : I \times \mathbb{R}^d \times L^2(\mathcal{P}_2(\mathbb{R}^d)) \rightarrow \mathbb{R}$. f is said to be convex if for every $u \in I$, $x, x' \in \mathbb{R}^d$, $\mu, \mu' \in L^2(\mathcal{P}_2(\mathbb{R}^d))$, we have :

$$f(u, x', \mu') - f(u, x, \mu) \geq \partial_x f(u, x, \mu) \cdot (x' - x) + \int_I \mathbb{E} \left[\partial_{\delta m} f(u, x, \mu)(\tilde{u}, X^{\tilde{u}}) \cdot (X'^{\tilde{u}} - X^{\tilde{u}}) \right] d\tilde{u}. \quad (9)$$

where $X'^{\tilde{u}} \sim \mu'^{\tilde{u}}$ and $X^{\tilde{u}} \sim \mu^{\tilde{u}}$.

Analysis tools on $L^2(\mathcal{P}_2(\mathbb{R}^d))$

A notion of convexity

A notion of convexity in $L^2(\mathcal{P}_2(\mathbb{R}^d))$

Let $f : I \times \mathbb{R}^d \times L^2(\mathcal{P}_2(\mathbb{R}^d)) \rightarrow \mathbb{R}$. f is said to be convex if for every $u \in I$, $x, x' \in \mathbb{R}^d$, $\mu, \mu' \in L^2(\mathcal{P}_2(\mathbb{R}^d))$, we have :

$$\begin{aligned} f(u, x', \mu') - f(u, x, \mu) &\geq \partial_x f(u, x, \mu) \cdot (x' - x) \\ &\quad + \int_I \mathbb{E} \left[\partial_{\delta m} f(u, x, \mu)(\tilde{u}, X^{\tilde{u}}) \cdot (X'^{\tilde{u}} - X^{\tilde{u}}) \right] d\tilde{u}. \end{aligned} \quad (9)$$

where $X'^{\tilde{u}} \sim \mu'^{\tilde{u}}$ and $X^{\tilde{u}} \sim \mu^{\tilde{u}}$.

- Can be extended to functions defined on $I \times \mathbb{R}^d \times L^2(\mathcal{P}_2(\mathbb{R}^d)) \times A$:

$$\begin{aligned} f(u, x', \mu', a') - f(u, x, \mu, a) &\geq \partial_x f(u, x, \mu, a) \cdot (a' - a) + \partial_a f(u, x, \mu, a) \cdot (a' - a) \\ &\quad + \int_I \mathbb{E} \left[\partial_{\delta m} f(u, x, \mu, a)(\tilde{u}, X^{\tilde{u}}) \cdot (X'^{\tilde{u}} - X^{\tilde{u}}) \right] d\tilde{u}. \end{aligned} \quad (10)$$

Controlled system \mathbf{X} dynamics

System dynamics for $\mathbf{X} = (X^u)_{u \in I}$:

$$\begin{cases} dX_t^u = b(\mathbf{u}, X_t^u, (\mathbb{P}_{X_t^v})_{v \in I}, \alpha_t^u) dt + \sigma(\mathbf{u}, X_t^u, (\mathbb{P}_{X_t^v})_{v \in I}, \alpha_t^u) dW_t^u & 0 \leq t \leq T, \\ X_0^u = \xi^u, u \in I. \end{cases} \quad (11)$$

with admissible controls $\alpha = (\alpha^u)_{u \in I}$ are defined as follows. For an arbitrary Borel measurable function $\alpha : I \times [0, T] \times \mathcal{C}_{[0, T]}^n \times (0, 1) \rightarrow A$, we define :

$$\alpha_t^u = \alpha(\mathbf{u}, t, W_{\cdot \wedge t}^u, U^u), \text{ and } \int_I \int_0^T \mathbb{E}[|\alpha_t^u|^2] dt du < +\infty. \quad (12)$$

Such α is said to be admissible and belongs to \mathcal{A} . Moreover, the initial condition $\xi = (\xi^u)_{u \in I}$ is an admissible initial condition if there exists a Borel measurable function $\xi : I \times (0, 1) \rightarrow \mathbb{R}^d$ s.t

$$\xi^u = \xi(\mathbf{u}, U^u), \text{ and } \int_I \mathbb{E}[|\xi^u|^2] du < +\infty. \quad (13)$$

Existence and uniqueness for **X**

Under some standard assumptions on model coefficients b, σ , for an admissible initial condition ξ and an admissible control $\alpha \in \mathcal{A}$, there exists a unique solution to (11) such that there exists a Borel measurable function x defined on $I \times \mathbb{R}^d \times \mathcal{C}_{[0,T]}^n \times (0,1)$ into \mathbb{R}^d with :

$$X_t^u = x(u, t, W_{\cdot \wedge t}^u, U^u), \quad \mathbb{P} \text{ a.s.}, \quad \forall (t, u) \in [0, T] \times I \text{ and } \int_I \mathbb{E} \left[\sup_{0 \leq t \leq T} |X_t^u|^2 \right] du < +\infty.$$

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→ This theorem implies the measurability of the mapping $u \mapsto \mathcal{L}(X^u, W^u, U^u)$ which implies under additional standard assumptions f and g that the cost functional (5) is well defined and finite.

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Some preliminaries tools

Definition of the Hamiltonian H

Definition of the Hamiltonian H

The Hamiltonian \mathbb{R} -valued function H of the stochastic optimization problem is defined as :

$$H(u, x, \mu, y, z, a) = b(u, x, \mu, a) \cdot y + \sigma(u, x, \mu, a) : z + f(u, x, \mu, a) \quad (14)$$

where $(u, x, \mu, y, z, a) \in I \times \mathbb{R}^d \times L^2(\mathcal{P}_2(\mathbb{R}^d)) \times \mathbb{R}^d \times \mathbb{R}^{d \times n} \times A$.

- Compute an optimality criterion involving the Hamiltonian H assuming differentiability and convexity as defined previously.
- In the following, A will denote a convex subset of \mathbb{R}^m for $m \in \mathbb{N}^*$.

Probabilistic set-up for non exchangeable mean field SDEs

Adjoint Equations to \mathbf{X}

We define the 2 following spaces :

$$L^2(I; \mathcal{S}^d) = \{\mathbf{Y} = (Y^u)_{u \in I} : Y^u \text{ is } \mathbb{F}^u\text{-adapted and } \int_I \mathbb{E} \left[\sup_{0 \leq t \leq T} |Y_t^u|^2 \right] du < +\infty\}$$

$$L^2(I; \mathbb{H}^{2,d \times n}) = \{\mathbf{Z} = (Z^u)_{u \in I} : Z^u \text{ is } \mathbb{F}^u\text{-adapted and } \int_I \mathbb{E} \left[\int_0^T |Z_t^u|^2 dt \right] du < +\infty\}$$

Adjoint Equations to \mathbf{X}

We call adjoint processes of \mathbf{X} any pair $(\mathbf{Y}, \mathbf{Z}) = (Y_t^u, Z_t^u)_{u \in I, t \in [0, T]}$ of processes in $L^2(I; \mathcal{S}^d) \times L^2(I; \mathbb{H}^{2,d \times n})$ satisfying the following conditions

(i) (\mathbf{Y}, \mathbf{Z}) is solution to the adjoint equations

$$\begin{cases} dY_t^u = -\partial_x H(u, X_t^u, (\mathbb{P}_{X_t^v})_{v \in I}, Y_t^u, Z_t^u, \alpha_t^u) dt + Z_t^u dW_t^u \\ \quad - \int_I \mathbb{E} \left[\partial_{\frac{\delta}{\delta m}} H(\tilde{u}, \tilde{X}_t^{\tilde{u}}, (\mathbb{P}_{X_t^v})_{v \in I}, \tilde{Y}_t^{\tilde{u}}, \tilde{Z}_t^{\tilde{u}}, \tilde{\alpha}_t^{\tilde{u}})(u, X_t^u) \right] d\tilde{u} dt, \quad t \in [0, T], \\ Y_T^u = \partial_x g(u, X_T^u, (\mathbb{P}_{X_T^v})_{v \in I}) + \int_I \mathbb{E} \left[\partial_{\frac{\delta}{\delta m}} g(\tilde{u}, \tilde{X}_T^{\tilde{u}}, (\mathbb{P}_{X_T^v})_{v \in I})(u, X_T^u) \right] d\tilde{u}, \end{cases} \quad (15)$$

for every $u \in I$ where $(\tilde{X}, \tilde{Y}, \tilde{Z}, \tilde{\alpha})$ is an independent copy of (X, Y, Z, α) defined on $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\mathbb{P}})$

(ii) There exist Borel functions y and z defined on $I \times [0, T] \times C_{[0, T]}^d \times (0, 1)$ such that

$$Y_t^u = y(u, t, W_{\cdot \wedge t}^u, U^u), \quad \text{and} \quad Z_t^u = z(u, t, W_{\cdot \wedge t}^u, U^u), \quad \text{for } t \in [0, T], \mathbb{P}\text{-a.s. and } u \in I.$$

Derivation of a Pontryagin Optimality Condition

A necessary condition

We now state the main results which are obtained under some standard regularity assumptions on b , σ , f and g .

Gâteaux derivative of J

For $\beta \in \mathcal{A}$ such that $\alpha + \epsilon\beta \in \mathcal{A}$ for $\epsilon > 0$ small enough, we have :

$$\lim_{\epsilon \rightarrow 0} \frac{1}{\epsilon} (J(\alpha + \epsilon\beta) - J(\alpha)) = \int_I \mathbb{E} \left[\int_0^T \left(\partial_\alpha H(\mathbf{u}, X_t^u, (\mathbb{P}_{X_t^u})_{v \in I}, Y_t^u, Z_t^u, \alpha_t^u) \cdot \beta_t^u \right) dt \right] du$$

where \mathbf{X} is given by (11), (\mathbf{Y}, \mathbf{Z}) are given by (15) and the Hamiltonian function H is given by (14).

Derivation of a Pontryagin Optimality Condition

A necessary condition

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where \mathbf{X} is given by (11), (\mathbf{Y}, \mathbf{Z}) are given by (15) and the Hamiltonian function H is given by (14).

Necessary condition for optimality of α

Moreover, if we assume that H is convex in $a \in A$, that $\alpha = (\alpha_t^u)_{u \in I, 0 \leq t \leq T}$ is optimal, that $\mathbf{X} = (X_t^u)_{u \in I, 0 \leq t \leq T}$ is the associated optimal control state given by (11) and that $(\mathbf{Y}, \mathbf{Z}) = (Y_t^u, Z_t^u)_{u \in I, 0 \leq t \leq T}$ are the associated adjoint processes solving (15), then we have for almost every $u \in I$:

$$\forall a \in A, \quad H(\mathbf{u}, X_t^u, (\mathbb{P}_{X_t^v})_{v \in I}, Y_t^u, Z_t^u, \alpha_t^u) \leq H(\mathbf{u}, X_t^u, (\mathbb{P}_{X_t^v})_{v \in I}, Y_t^u, Z_t^u, a) \quad dt \otimes d\mathbb{P} \text{ a.e.} \quad (16)$$

Derivation of a Pontryagin Optimality Condition

A sufficient condition

Sufficient condition for optimality of α

Let $\alpha = (\alpha^u)_{u \in I} \in \mathcal{A}$, \mathbf{X} the corresponding controlled state process and (\mathbf{Y}, \mathbf{Z}) the corresponding adjoint processes. Let also assume that for almost every $u \in I$:

- (1) $\mathbb{R}^d \times L^2(\mathcal{P}_2(\mathbb{R}^d)) \ni (x, \mu) \rightarrow g(u, x, \mu)$ is convex
- (2) $\mathbb{R}^d \times L^2(\mathcal{P}_2(\mathbb{R}^d)) \times A \ni (x, \mu, a) \rightarrow H(u, x, \mu, Y_t^u, Z_t^u, a)$ is convex $dt \otimes d\mathbb{P}$ a.e

If we assume also following the necessary condition for optimality that for almost every $u \in I$:

$$H(u, X_t^u, (\mathbb{P}_{X_t^v})_{v \in I}, Y_t^u, Z_t^u, \alpha_t^u) = \inf_{\beta \in A} H(u, X_t^u, (\mathbb{P}_{X_t^v})_{v \in I}, Y_t^u, Z_t^u, \beta), \quad dt \otimes d\mathbb{P} \text{ a.e}$$

Then, α is an optimal control in the sense that $J(\alpha) = \inf_{\alpha' \in \mathcal{A}} J(\alpha')$

- Recall that the convexity property is understood under the definition (9).

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Solvability of the collection of FBSDE

Definition of a solution

The Pontryagin Maximum principle leads us to study the following collection of fully coupled FBSDE :

Collection of FBSDE system

$$\left\{ \begin{array}{l} dX_t^u = b(u, t, X_t^u, (\mathbb{P}_{X_t^v})_{v \in I}, \hat{\alpha}_t^u) dt + \sigma(u, t, X_t^u, (\mathbb{P}_{X_t^v})_{v \in I}, \hat{\alpha}_t^u) dW_t^u, \\ X_0^u = \xi^u, \\ dY_t^u = -\partial_x H(u, t, X_t^u, (\mathbb{P}_{X_t^v})_{v \in I}, Y_t^u, Z_t^u, \hat{\alpha}_t^u) dt + Z_t^u dW_t^u \\ \quad - \int_I \tilde{\mathbb{E}} \left[\partial \frac{\delta}{\delta m} H(\tilde{u}, t, \tilde{X}_t^{\tilde{u}}, (\mathbb{P}_{X_t^v})_{v \in I}, \tilde{Y}_t^{\tilde{u}}, \tilde{Z}_t^{\tilde{u}}, \tilde{\alpha}_t^{\tilde{u}})(u, X_t^u) \right] d\tilde{u} dt, \\ Y_T^u = \partial_x g(u, X_T^u, (\mathbb{P}_{X_T^v})_{v \in I}) + \int_I \tilde{\mathbb{E}} \left[\partial \frac{\delta}{\delta m} g(\tilde{u}, \tilde{X}_T^{\tilde{u}}, (\mathbb{P}_{X_T^v})_{v \in I})(u, X_T^u) \right] d\tilde{u}, \\ \hat{\alpha}_t^u = \hat{\alpha}(u, t, X_t^u, (\mathbb{P}_{X_t^v})_{v \in I}, Y_t^u, Z_t^u), \end{array} \right. \quad (17)$$

for every $u \in I$, $t \in [0, T]$. $(\tilde{X}, \tilde{Y}, \tilde{Z}, \tilde{\alpha})$ is an independant copy of (X, Y, Z, α) defined on $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\mathbb{P}})$ and $\tilde{\mathbb{E}}$ denotes the expectation on the probability space $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\mathbb{P}})$

Solvability of the collection of FBSDE

Definition of a solution

The Pontryagin Maximum principle leads us to study the following collection of fully coupled FBSDE :

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→ Note that the previous system is indeed a fully coupled FBSDE through the definition of $\hat{\alpha}_t^u$.

→ We are looking for a solution to (17) in a sense to be defined.

Solvability of a collection of BSDEs

Definition of a solution

Definition of a suitable space of solution \mathcal{S} to the collection of FBSDE (17)

We say that $(X, Y, Z) = (X^u, Y^u, Z^u)_{u \in I}$ belongs to \mathcal{S} if :

- There exists measurable functions x, y and z defined on $I \times [0, T] \times C_{[0, T]}^d \times [0, 1] \rightarrow \mathbb{R}^d$ such that

$$X_t^u = x(u, t, W_{\cdot \wedge t}^u, U^u), \quad Y_t^u = y(u, t, W_{\cdot \wedge t}^u, U^u), \quad \text{and} \quad Z_t^u = z(u, t, W_{\cdot \wedge t}^u, U^u). \quad (18)$$

- Each process X^u and Y^u are \mathbb{F}^u -adapted and continuous and Z^u is \mathbb{F}^u -adapted and square integrable.
- The following norm is finite :

$$\|(X, Y, Z)\|_{\mathcal{S}} = \left(\int_I \mathbb{E} \left[\sup_{t \in [0, T]} |X_t^u|^2 + \sup_{t \in [0, T]} |Y_t^u|^2 + \int_0^T |Z_t^u|^2 dt \right] du \right)^{\frac{1}{2}}$$

We say that $(X^u, Y^u, Z^u)_{u \in I}$ is a unique solution to (17) if the equations in (17) are satisfied for almost every u . Moreover, we say that the solution is unique if, whenever $(X^u, Y^u, Z^u)_{u \in I}$, $(\tilde{X}^u, \tilde{Y}^u, \tilde{Z}^u)_{u \in I}$, the processes (X^u, Y^u, Z^u) and $(\tilde{X}^u, \tilde{Y}^u, \tilde{Z}^u)$ coincide, up to a \mathbb{P} -null set, for almost every $u \in I$.

→ Note that (18) guarantees the measurability of the mapping

$$I \ni u \mapsto \mathcal{L}(X^u, Y^u, Z^u) \in \mathcal{P}_2(C_{[0, T]}^d \times C_{[0, T]}^d \times \mathbb{H}_{[0, T]}^{2, d \times n})$$

which justifies the well defined norm $\|\cdot\|_{\mathcal{S}}$.

Solvability of a collection of FBSDEs

Assumptions : Existence and Uniqueness

We now give some assumptions which will ensure existence and uniqueness to the collection of FBSDE (17).

Assumption : Existence and Uniqueness (1)

There exists two constants $L \geq 0$ and $\lambda > 0$ such that :

- (i) The drift b and the volatility σ are linear in μ , x and a such that :

$$\begin{aligned} b(u, x, \mu, a) &= b_0(u) + \int_I b_2(u, v) \tilde{\mu}^v dv + b_3(u)x + b_4(u)a \\ \sigma(u, x, \mu, a) &= \sigma_0(u) + \int_I \sigma_2(u, v) \tilde{\mu}^v dv + \sigma_3(u)x + \sigma_4(u)a \end{aligned}$$

for some bounded measurable deterministic functions b_0, b_1, b_2, b_3, b_4 with values in $\mathbb{R}^d, \mathbb{R}^{d \times d}, \mathbb{R}^{d \times d}, \mathbb{R}^{d \times m}$ and $\sigma_0, \sigma_1, \sigma_2, \sigma_3$ with values in $\mathbb{R}^{d \times n}, \mathbb{R}^{(d \times n) \times d}, \mathbb{R}^{(d \times n) \times d}$ and $\mathbb{R}^{(d \times n) \times m}$ and where the notation $\tilde{\mu}^v = \int_{\mathbb{R}^d} x \mu^v(dx)$.

- (ii) The functions f and g satisfy the same assumptions as previously. Moreover, the derivatives of f and g with respect to (x, a) and x respectively are assumed to be L -Lipschitz with respect to (x, a, μ) and (x, μ) respectively where the Lipschitz property in the variable μ is understood in the sense of the distance (6).
- (iii) For any $u \in I$, any $x, x' \in \mathbb{R}^d$, any $a, a' \in A$ any $\mu = (\mu^u)_{u \in I}, \mu' = (\mu'^u)_{u \in I} \in L^2(\mathcal{P}_2(\mathbb{R}^d))$, and any \mathbb{R}^d random variables X^u and X'^u such that $X^u \sim \mu^u$ and $X'^u \sim \mu'^u$, we have :

$$\begin{aligned} &\int_I \mathbb{E} \left[\left| \partial_{\delta m} \frac{\delta}{\delta m} f(u, x', \mu', a')(\tilde{u}, X'^{\tilde{u}}) - \partial_{\delta m} \frac{\delta}{\delta m} f(u, x, \mu, a)(\tilde{u}, X^{\tilde{u}}) \right|^2 \right] d\tilde{u} \\ &\leq L \left(|x' - x|^2 + |a' - a|^2 + \int_I \mathbb{E}[|X'^u - X^u|^2] du \right) \end{aligned}$$

Solvability of a collection of FBSDEs

Assumptions : Existence and Uniqueness

Assumption : Existence and Uniqueness (2)

Similarly for g , we have :

$$\begin{aligned} \int_I \mathbb{E} \left[\left| \partial \frac{\delta}{\delta m} g(u, x', \mu')(\tilde{u}, X^{\prime \tilde{u}}) - \partial \frac{\delta}{\delta m} g(u, x, \mu)(\tilde{u}, X^{\tilde{u}}) \right|^2 \right] d\tilde{u} \\ \leq L \left(|x' - x|^2 + \int_I \mathbb{E}[|X'^u - X^u|^2] du \right) \end{aligned}$$

(iv) The function f satisfies the following convexity property :

$$\begin{aligned} f(u, x', \mu', a') - f(u, x, \mu, a) - \partial_x f(u, x, \mu, a) \cdot (x' - x) - \partial_a f(u, x, \mu, a) \cdot (a' - a) \\ - \int_I \mathbb{E} \left[\partial \frac{\delta}{\delta m} f(u, x, \mu, a)(\tilde{u}, X^{\tilde{u}}) \cdot (X^{\prime \tilde{u}} - X^{\tilde{u}}) \right] d\tilde{u} \geq \lambda |a' - a|^2 \end{aligned}$$

for all $u \in I$, $(x, \mu, a) \in \mathbb{R}^d \times L^2(\mathcal{P}_2(\mathbb{R}^d)) \times A$ and $(x', \mu', a') \in \mathbb{R}^d \times L^2(\mathcal{P}_2(\mathbb{R}^d)) \times A$, when $X^{\tilde{u}} \sim \mu^{\tilde{u}}$ and $X^{\prime \tilde{u}} \sim \mu^{\prime \tilde{u}}$. We also assume that g is convex in (x, μ) as we did in the sufficient condition in the Pontryagin optimality principle :

$$g(u, x', \mu') - g(u, x, \mu) - \partial_x g(u, x, \mu) \cdot (x' - x) - \int_I \mathbb{E} \left[\partial \frac{\delta}{\delta m} g(u, x, \mu)(\tilde{u}, X^{\tilde{u}}) \cdot (X^{\prime \tilde{u}} - X^{\tilde{u}}) \right] d\tilde{u} \geq 0$$

Solvability of a collection of FBSDEs

Existence and unicity

Theorem : Existence and Uniqueness for a solution to (22)

Under Assumptions 23 and 24 and for any admissible initial condition $\xi = (\xi^u)_{u \in I}$, the collection of forward backward system $(\mathbf{X}, \mathbf{Y}, \mathbf{Z})$ (22) is uniquely solvable in \mathcal{S} .

→ The proof is based on a classical method of continuation similarly as done in the book of Carmona and Delarue (see [4]).

- 1 Introduction
- 2 Some preliminaries tools on $L^2(\mathcal{P}_2(\mathbb{R}^d))$
- 3 Pontryagin principle for optimality
- 4 Solvability of non exchangeable mean field FBSDEs
- 5 An application to linear quadratic mean field control

Linear Quadratic (LQ) graphon models

Model introduction

Linear graphon models

We consider the following class of models (assuming for sake of simplicity σ constant and $A = \mathbb{R}^m$).

$$\begin{aligned} dX_t^u &= \left[\beta^u + A^u X_t^u + \int_I G_A(u, v) \mathbb{E}[X_t^v] dv + B^u \alpha_t^u \right] dt + \gamma^u dW_t^u, t \in [0, T] \\ X_0^u &= \xi^u, u \in I, \end{aligned} \quad (19)$$

where $\xi = (\xi^u)_u$ an admissible initial condition and $\beta \in L^\infty(I; \mathbb{R}^d)$, $\gamma \in L^\infty(I; \mathbb{R}^d)$, $A \in L^\infty(I, \mathbb{R}^{d \times d})$, $B \in L^\infty(I; \mathbb{R}^{d \times m})$, $G_A \in L^\infty(I \times I; \mathbb{R}^{d \times d})$.

Quadratic cost functional

$$\begin{aligned} J(\alpha) &= \int_I \mathbb{E} \left[\int_0^T Q^u(X_t^u - \int_I \tilde{G}_Q(u, v) \mathbb{E}[X_t^v] dv) \cdot (X_t^u - \int_I \tilde{G}_Q(u, v) \mathbb{E}[X_t^v] dv) + \alpha_t^u \cdot R^u \alpha_t^u \right. \\ &\quad + 2\alpha_t^u \cdot \Gamma^u X_t^u + 2\alpha_t^u \cdot \int_I G_I(u, v) \mathbb{E}[X_t^v] dv dt \\ &\quad \left. + H^u(X_T^u - \int_I \tilde{G}_H(u, v) \mathbb{E}[X_T^v] dv) \cdot (X_T^u - \int_I \tilde{G}_H(u, v) \mathbb{E}[X_T^v] dv) \right] du \end{aligned} \quad (20)$$

Ansatz form for \mathbf{Y}

We are looking for an ansatz Y_t^u in the following form :

$$Y_t^u = 2(K^u(t)X_t^u + \int_I \bar{K}_t(u, v)\mathbb{E}[X_t^v]dv) + \Lambda_t^u, \quad (21)$$

where $K \in C^1([0, T]; L^\infty(I; \mathbb{S}_+^d))$, $\bar{K} \in C^1([0, T], L^2(I \times I; \mathbb{R}^{d \times d}))$ and $\Lambda \in C^1([0, T]; L^2(I; \mathbb{R}^d))$ are to be determined through infinite dimensional Riccati equations.

→ We inject the form (21) in (17) and we end up with a triangular Riccati system for K , \bar{K} and Λ for which we can prove existence and uniqueness.

→ Finally, we can show existence and uniqueness of the following collection of SDE :

$$\begin{cases} dX_t^u = \left(\beta^u - B^u(R^u)^{-1}(B^u)^\top \Lambda_t^u + \left(A^u - B^u(R^u)^{-1}((B^u)^\top K_t^u + \Gamma^u) \right) X_t^u \right. \\ \quad \left. + \int_I \left(G_A(u, v) - B^u(R^u)^{-1}((B^u)^\top \bar{K}_t(u, v) + G_I(u, v)) \right) \mathbb{E}[X_t^v] dv \right) dt + \gamma^u dW_t^u, \\ X_0^u = \xi^u, \end{cases}$$

Optimal control α in the L-Q case

$$\alpha_t^u = S^u(t)X_t^u + \int_I \bar{S}^{uv}(t)\mathbb{E}[X_t^v]dv + \Gamma^u(t), \quad (22)$$

where $\mathbf{S} = (S^u)_u$, $\bar{\mathbf{S}} = (\bar{S}^{uv})_{u,v}$ and $\mathbf{\Gamma} = (\Gamma^u)_u$ are deterministic functions, expressed in terms of K, \bar{K} and Λ given by :

$$\begin{cases} S^u(t) = -(R^u)^{-1} \left((B^u)^\top K_t^u + \Gamma^u \right), \\ \bar{S}^{uv}(t) = -(R^u)^{-1} \left((B^u)^\top \bar{K}_t(u, v) + G_I(u, v) \right) \\ \Gamma^u(t) = -(R^u)^{-1} (B^u)^\top \Lambda_t^u \end{cases}$$

Linear Quadratic Graphon Mean Field Control

An example : systemic risk model with heterogeneous banks

We consider the following model :

$$\begin{aligned}dX_t^u &= [\kappa(X_t^u - \int_I \tilde{G}_\kappa(u, v) \mathbb{E}[X_t^v] dv) + \alpha_t^u] dt + \sigma^u dW_t^u, \\X_0^u &= \xi^u,\end{aligned}\tag{23}$$

with \tilde{G}_κ a bounded, symmetric measurable function from $I \times I$ into \mathbb{R} ., $\sigma^u > 0$ and $\alpha = (\alpha^u)$ the control process. The initial condition $\xi = (\xi^u)_u$ is assumed to be admissible. The aim of the central bank is then to minimize over α the following cost functional :

Cost functional $J(\alpha)$

$$\begin{aligned}J(\alpha) &= \int_I \mathbb{E} \left[\int_0^T \left[\eta^u (X_t^u - \int_I \tilde{G}_\eta(u, v) \mathbb{E}[X_t^v] dv)^2 + q^u \alpha_t^u (X_t^u - \int_I G_q(u, v) \mathbb{E}[X_t^v] dv) + |\alpha_t^u|^2 \right] dt \right. \\&\quad \left. + r^u (X_T^u - \int_I \tilde{G}_r(u, v) \mathbb{E}[X_T^v] dv)^2 \right] du\end{aligned}\tag{24}$$

Linear Quadratic Graphon Mean Field Control

An example : systemic risk model with heterogeneous banks

Optimal control form in systemic risk model

Following (22), we end up with the following optimal control form :

$$\hat{\alpha}_t^u = -(\mathcal{K}_t^u + \frac{q^u}{2})(X_t^u - \int_I G_Q(u, v) \mathbb{E}[X_t^v] dv) - \int_I (\bar{K}_t(u, v) + \mathcal{K}_t^u G_Q(u, v)) \mathbb{E}[X_t^v] dv \quad (25)$$

• Setting the coefficients independent of $u \in I$ and $\tilde{G}_\kappa \equiv \tilde{G}_\eta \equiv \tilde{G}_r \equiv G_q \equiv 1$, we recover the classical mean-field result see for systemic risk with $\bar{K}_t \equiv -K_t$ and the optimal control is given by :

$$\hat{\alpha}_t = -(K_t + \frac{q}{2})(X_t - \mathbb{E}[X_t])$$

→ We therefore have the additional term $\int_I (\bar{K}_t(u, v) + \mathcal{K}_t^u G_Q(u, v)) \mathbb{E}[X_t^v] dv$ compared to the homogeneous case.

Conclusion

Main results of our work

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- We provide a natural extension to the classical MFC problem in the context of non exchangeable interactions by considering the space $L^2(\mathcal{P}_2(\mathbb{R}^d))$.

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- We provide a natural extension to the classical MFC problem in the context of non exchangeable interactions by considering the space $L^2(\mathcal{P}_2(\mathbb{R}^d))$.
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- We provide a natural extension to the classical MFC problem in the context of non exchangeable interactions by considering the space $L^2(\mathcal{P}_2(\mathbb{R}^d))$.
- It leads to the study of a collection indexed by $u \in I$ of fully coupled FBSDE for which we are able to prove existence and uniqueness under standard assumptions on the model coefficients.
- We provide a semi-analytic form of the LQ graphon model as it leads to the study of infinite dimensional rectangular Riccati equations.

Conclusion

Further works on *NE-MFC* problem

Future works on *NE – MFC*

- Study of the convergence of the N -agent system towards the limit candidate either by showing convergence of value functions of both problems (**weak formulation**) or convergence of optimal controls (**strong formulation**). (Partially done in the context of graphons based interactions in the work of Cao and Laurière).
- Study of the *NE – MFC* with common noise.
- Numerical algorithms in the context of a finite number of players :
 - (1) In a model-based setting : Learning optimal controls $\alpha = (\alpha^{1,N}, \dots, \alpha^{N,N})$ and value function V_N through Deep Learning algorithms.
 - (2) In a model-free setting : Learning optimal controls $\alpha = (\alpha^{1,N}, \dots, \alpha^{N,N})$ and value function V_N through Reinforcement Learning algorithms.



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