

On the approximation of first order mean field games

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Introduction: a static symmetric game

Let $Q \subseteq \mathbb{R}^d$ be compact and $F : Q \times \mathcal{P}(Q) \rightarrow \mathbb{R}$ be continuous. We consider a game defined by

- ▶ N players.
- ▶ Set of actions Q (the same for all the players).
- ▶ A cost functional $F_i : Q^N \rightarrow \mathbb{R}$ for Player i defined by

$$F_i(x_1, \dots, x_i, \dots, x_N) = F \left(x_i, \frac{1}{N-1} \sum_{j \neq i} \delta_{x_j} \right).$$

Example: $N =$ number of swimmers, Q is a beach and

$$F \left(x_i, \frac{1}{N-1} \sum_{j \neq i} \delta_{x_j} \right) = \alpha d(x_i, \text{snack bar}) - \frac{\beta}{N-1} \sum_{j \neq i} |x_i - x_j|$$

In order to obtain the existence of Nash equilibria, we consider the game with mixed strategies, i.e.

- ▶ The new set of actions is $\mathcal{P}(Q)$ (the same for all the players).
- ▶ The new cost functional $F_{rel,i} : \mathcal{P}(Q)^N \rightarrow \mathbb{R}$ for Player i defined by

$$F_{rel,i}(m_1, \dots, m_i, \dots, m_N) = \int_{Q^N} F_i(x_1, \dots, x_i, \dots, x_N) \otimes_{j=1}^N dm_j(x_j).$$

A configuration $(\bar{m}_1, \dots, \bar{m}_N)$ is a Nash equilibrium if $\forall i = 1, \dots, N$

$$F_{rel,i}(\bar{m}_1, \dots, \bar{m}_i, \dots, \bar{m}_N) \leq F_{rel,i}(\bar{m}_1, \dots, m, \dots, \bar{m}_N) \quad \forall m \in \mathcal{P}(Q).$$

This relaxed framework and the symmetry of the game allow to show the existence of at least one equilibrium having the form

$$(\bar{m}^N, \dots, \bar{m}^N) \quad \text{for some } \bar{m}^N \in \mathcal{P}(Q).$$

- ▶ The crucial point here is that, as $N \rightarrow \infty$, any limit point m^* of m_N satisfies

$$\int_Q F(x, m^*) dm^*(x) = \min_{m \in \mathcal{P}(Q)} \int_Q F(x, m) dm(x),$$

or, equivalently,

$$\text{supp}(m^*) \subseteq \text{argmin} \{F(x, m^*) \mid x \in Q\}.$$

Consequently, any of the two previous relations can be used to define the notion of MFG equilibrium.

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The dynamic and deterministic case (based on a joint work with M. Fischer)

A model problem in continuous time:

- ▶ Consider two continuous functions $f, g : \mathbb{R}^d \times \mathcal{P}_1(\mathbb{R}^d) \rightarrow \mathbb{R}$, differentiable w.r.t. the first variable and satisfying that

$$\sup_{m \in \mathcal{P}_1(\mathbb{R}^d)} \{ \|f(\cdot, m)\|_{C^1} + \|g(\cdot, m)\|_{C^1} \} \leq C,$$

- ▶ Consider N **players**, positioned at $x_1, \dots, x_N \in \mathbb{R}^d$ at time $t = 0$.
- ▶ The set of actions for Player i is $\mathcal{A}(x_i)$, where

$$(\forall x \in \mathbb{R}^d) \quad \mathcal{A}(x) := \{ \gamma \in H^1([0, T]; \mathbb{R}^d) \mid \gamma(0) = x \}.$$

- ▶ Given $\gamma_1 \in \mathcal{A}(x_1), \dots, \gamma_N \in \mathcal{A}(x_N)$, the cost for Player i is

$$j_i(\gamma_1, \dots, \gamma_i, \dots, \gamma_N) = j \left(\gamma_i, \frac{1}{N-1} \sum_{j \neq i} \delta_{\gamma_j} \right),$$

where, setting $\Gamma = C([0, T]; \mathbb{R}^d)$, $j : H^1([0, T]; \mathbb{R}^d) \times \mathcal{P}(\Gamma) \rightarrow \mathbb{R}$ is given by

$$j(\gamma_i, m) := \int_0^T \left[\frac{1}{2} |\dot{\gamma}_i(t)|^2 + f(\gamma_i(t), m(t)) \right] dt + g(\gamma_i(T), m(T)),$$

with

$$m(t) := e(t) \# m \quad \forall t \in [0, T].$$

Notice that

$$e(t) \# \left(\frac{1}{N-1} \sum_{j \neq i} \delta_{\gamma_j} \right) = \frac{1}{N-1} \sum_{j \neq i} \delta_{\gamma_j(t)}.$$

The previous game is not symmetric due to the heterogeneity of the initial conditions. In order to obtain a symmetric game let us

- ▶ assume that the initial positions of the players are randomly, identically and independently distributed.
- ▶ We denote by m_0 their common initial distribution, which is assumed to have a **compact support**.

- ▶ In this context, it is natural to define the action set of the players as

$$\mathcal{A} := \{\gamma : \mathbb{R}^d \rightarrow \Gamma \mid \gamma \text{ is Borel meas. and}$$

$$\gamma^x := \gamma(x) \in \mathcal{A}(x), \quad \forall x \in \text{supp}(m_0)\}.$$

- ▶ Accordingly, given the strategies $\gamma_1 \in \mathcal{A}, \dots, \gamma_N \in \mathcal{A}$, the cost functional for Player i is redefined as

$$J_i(\gamma_1, \dots, \gamma_i, \dots, \gamma_N) = J(\gamma_i, (\gamma_j)_{j \neq i}),$$

where

$$J(\gamma_i, (\gamma_j)_{j \neq i}) := \int_{(\mathbb{R}^d)^N} j\left(\gamma_i^{x_i}, \frac{1}{N-1} \sum_{j \neq i} \delta_{\gamma_j^{x_j}}\right) \otimes_{j=1}^N dm_0(x_j).$$

- ▶ Set $m_j := \gamma_j \# m_0$ (i.e. $dm_j(\gamma) = d\delta_{\gamma_j^x}(\gamma) dm_0(x)$). Then

$$J(\gamma_i, (\gamma_j)_{j \neq i}) = \int_{\Gamma^N} j\left(\gamma'_i, \frac{1}{N-1} \sum_{j \neq i} \delta_{\gamma'_j}\right) \otimes_{j=1}^N dm_j(\gamma'_j).$$

- ▶ If $(\bar{\gamma}_1, \dots, \bar{\gamma}_i, \dots, \bar{\gamma}_N)$ is a Nash equilibrium for the previous game, then there exists $C > 0$, independent of N , such that

$$(\forall x \in \text{supp}(m_0)) \quad \|\dot{\bar{\gamma}}^x\|_\infty \leq C.$$

- ▶ Set $Q_C := \{\gamma \in W^{1,\infty}([0, T]; \mathbb{R}^d) \mid \|\dot{\gamma}\|_\infty \leq C, \gamma(0) \in \text{supp}(m_0)\}$.
- ▶ It is natural to consider as set of strategies the **compact set**

$$A_{rel} := \{m \in \mathcal{P}(\Gamma) \mid e_0 \# m = m_0, \text{supp}(m) \subseteq Q_C\}.$$

and, as cost functional for Player i ,

$$J_{rel,i}(m_1, \dots, m_i, \dots, m_N) = J_{rel}(m_i, (m_j)_{j \neq i}), \quad \text{with}$$

$$J_{rel}(m_i, (m_j)_{j \neq i}) = \int_{\Gamma^N} j \left(\gamma_i, \frac{1}{N-1} \sum_{j \neq i} \delta_{\gamma_j} \right) \otimes_{j=1}^N dm_j(\gamma_j).$$

- ▶ Then standard techniques show the existence of a Nash **equilibrium** for the previous game having the form $(\bar{m}^N, \dots, \bar{m}^N)$.

Theorem: As $N \rightarrow \infty$, every limit point m^* of $(\bar{m}^N)_{N \in \mathbb{N}}$ satisfies

$$\int_{\Gamma} j(\gamma, m^*) dm^*(\gamma) = \min_{m \in \mathcal{A}_{rel}} \int_{\Gamma} j(\gamma, m) dm(\gamma),$$

or, equivalently,

$$\text{supp}(m^*) \subseteq \{\gamma \in \Gamma \mid \gamma \in \text{argmin}\{j(\gamma', m^*) \mid \gamma'(0) = \gamma(0)\}\}. \quad (*)$$

- ▶ A measure $m^* \in \mathcal{A}_{rel}$ is called a **mean field game equilibrium** if it satisfies $(*)$.
- ▶ The previous analysis shows, in particular, the existence of a MFG equilibrium.

- ▶ Suppose that in addition we have

$$\sup_{m \in \mathcal{P}_1(\mathbb{R}^d)} \{ \|f(\cdot, m)\|_{C^2} + \|g(\cdot, m)\|_{C^2} \} \leq C,$$

and that m_0 is **absolutely continuous** w.r.t. to \mathcal{L}^d .

- ▶ Then, associated to a MFG equilibrium m^* , there exists $\gamma^* \in \mathcal{A}$ such that $m^* = \gamma^* \# m_0$.
- ▶ Set $\rho(t) = m^*(t)$ and define the value function

$$v(t, x) = \inf \left\{ \int_t^T \left[\frac{1}{2} |\dot{\gamma}(s)|^2 + f(\gamma(s), \rho(s)) \right] ds + g(\gamma(T), \rho(T)) \mid \gamma \in H^1([t, T]; \mathbb{R}^d), \gamma(t) = x \right\}.$$

- ▶ Then the couple (v, ρ) solves

$$\left. \begin{aligned} -\partial_t v + \frac{1}{2} |\nabla v|^2 &= f(x, \rho(t)), & v(\cdot, T) &= g(\cdot, \rho(T)), \\ \partial_t \rho - \operatorname{div}(\nabla v \rho) &= 0, & \rho(\cdot, 0) &= m_0. \end{aligned} \right\} \quad (\text{MFG})$$

- ▶ Under the same assumptions, we also have the existence of $\bar{\gamma}^N \in \mathcal{A}$ such that $\bar{m}^N = \bar{\gamma}^N \# m_0$.
- ▶ As in the limit case, \bar{m}^N can be characterized by the solution (v^N, ρ^N) of a PDE system similar system to (MFG).
- ▶ This allows a convergence proof based on PDE techniques.

- ▶ **Uniqueness** of a solution to (MFG) holds if, for $h = f, g$,

$$\int_Q (h(x, \mu) - h(x, \mu')) d(\mu - \mu')(x) \geq 0, \quad \forall \mu, \mu' \in \mathcal{P}_1(\mathbb{R}^d).$$

- ▶ Related works:

Existence of MFG equilibria in the deterministic case: Lasry-Lions '07, Cannarsa-Capuani '18, Cannarsa-Capuani-Cardaliaguet '18, Mazanti-Santambrogio '18, Achdou-Mannucci-Marchi-Tchou '19, Cannarsa-Mendico '19,...

Convergence result: Lacker '16, Fischer '17, Cardaliaguet-Delarue-Lasry-Lions '19, Lacker '20, Gangbo-Mészáros '20,...

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Approximation of deterministic mean field games (based on a joint work with S. Hadikhanloo and an ongoing work with J. Gianatti)

Consider the MFG system

$$\left. \begin{aligned} -\partial_t v + \frac{1}{2} |\nabla v|^2 &= f(x, \rho(t)) && \text{in } [0, T] \times \mathbb{R}^d \\ v(T, \cdot) &= g(\cdot, \rho(T)) && \text{in } \mathbb{R}^d \\ \partial_t \rho - \operatorname{div}(\nabla v \rho) &= 0 && \text{in } [0, T] \times \mathbb{R}^d \\ \rho(0, \cdot) &= m_0 && \text{in } \mathbb{R}^d. \end{aligned} \right\} \quad (\text{MFG})$$

- ▶ A semi-Lagrangian scheme to solve (MFG) has been proposed in Carlini-S. '14. Full-convergence result when $d = 1$.
- ▶ Fourier methods to treat (MFG) have been proposed recently in Nuberkyan-Saúde'19 and Li-Jacobs-Li-Nuberkyan-Osher '20.
- ▶ We describe now an fully-discrete scheme, which approximate general MFG equilibria in the form (*).

- ▶ The discretization in Carlini-S.'14 is mainly based on the representation formulae

$$v(t, x) = \inf \int_t^T \left[\frac{1}{2} |\alpha(s)|^2 + f(\gamma(s), m(s)) \right] ds + g(\gamma(T), m(T))$$

$$\text{s.t.} \quad \dot{\gamma}(s) = \alpha(s) \quad \text{in } (t, T), \quad \gamma(t) = x$$

$$= \inf \int_t^T \left[\frac{1}{2} |\dot{\gamma}(s)|^2 + f(\gamma(s), m(s)) \right] ds + g(\gamma(T), m(T))$$

$$\text{s.t.} \quad \gamma(t) = x,$$

and

$$\rho(t) = \gamma^{(\cdot)}(t) \# m_0,$$

where, for $x \in \mathbb{R}^d$, $\gamma^x(t)$ is the solution, evaluated at time t , to

$$\dot{\gamma}(s) = -\nabla v(s, \gamma(s)) \quad \text{in } (0, T), \quad \gamma(0) = x.$$

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Approximation of optimal control problem solved by the typical agent

At the equilibrium, a typical agent solves a problem having the form

$$\inf \int_0^T \left[\frac{1}{2} |\alpha(t)|^2 + f(\gamma(t)) \right] dt + g(\gamma(T)) \quad \text{s.t.} \quad \dot{\gamma}(t) = \alpha(t), \quad \gamma(0) = x.$$

The associated HJB equation is given by

$$\begin{aligned} -\partial_t v + \frac{1}{2} |\nabla v|^2 &= f & \text{in } [0, T] \times \mathbb{R}^d \\ v(T, \cdot) &= g & \text{in } \mathbb{R}^d \end{aligned}$$

- Let $\Delta t > 0$, set $t_k = k(\Delta t)$ and $\mathcal{T}_{\Delta t} := \{0, t_1, \dots, t_n\}$, with $t_n = T$. A standard semi-discrete scheme to approximate v is

$$\begin{aligned} v_{\Delta t}(t_k, x) &= \inf_{\alpha \in \mathbb{R}^d} \left\{ \Delta t \left[\frac{|\alpha|^2}{2} + f(x) \right] + v_{\Delta t}(t_{k+1}, x + \Delta t \alpha) \right\}, \\ v_{\Delta t}(T, x) &= g(x). \end{aligned}$$

- ▶ Given a space-step $\Delta x > 0$, set $\mathcal{G}_{\Delta x} = \{x_i = i(\Delta x) \mid i \in \mathbb{Z}^d\}$. The fully-discrete SL scheme is

$$\begin{aligned} v_{\Delta t, \Delta x}(t_k, x_i) &= \inf_{\alpha \in \mathbb{R}^d} \left\{ \Delta t \left[\frac{|\alpha|^2}{2} + f(x_i) \right] + I[v_{\Delta t, \Delta x}](t_{k+1}, x_i + \Delta t \alpha) \right\}, \\ v_{\Delta t, \Delta x}(T, x_i) &= g(x_i), \end{aligned}$$

where $I[\cdot]$ is an interpolation operator associated to a triangulation with vertices in $\mathcal{G}_{\Delta x}$.

- ▶ Given the particular structure of the dynamics, we can avoid an infinite grid and, more importantly, interpolation by choosing controls such that

$$x_i + \Delta t \alpha \text{ is a grid point.}$$

Moreover, since the optimal control for the continuous problem will be bounded by some $C > 0$, it is natural to impose

$$\alpha = \frac{x_j - x_i}{\Delta t}, \quad |\alpha| \leq C.$$

- Setting $\mathcal{G}_{\Delta x}(x_i) = \{x_j \in \mathcal{G}_{\Delta x} \mid |x_j - x_i| \leq C\Delta t\}$, we get

$$v_{\Delta t, \Delta x}(t_k, x_i) = \inf_{x_j \in \mathcal{G}_{\Delta x}(x_i)} \left\{ \Delta t \left[\frac{1}{2} \left| \frac{x_j - x_i}{\Delta t} \right|^2 + f(x_i) \right] + v_{\Delta t, \Delta x}(t_{k+1}, x_j) \right\}$$

$$v_{\Delta t, \Delta x}(T, x_i) = g(x_i),$$

or, equivalently,

$$v_{\Delta t, \Delta x}(t_k, x_i) = \inf_{p \in \mathcal{P}(\mathcal{G}_{\Delta x}(x_i))} \left\{ \sum_{x_j \in \mathcal{G}_{\Delta x}(x_i)} p_j \left[\Delta t \left[\frac{1}{2} \left| \frac{x_j - x_i}{\Delta t} \right|^2 + f(x_i) \right] + v_{\Delta t, \Delta x}(t_{k+1}, x_j) \right] \right\},$$

$$v_{\Delta t, \Delta x}(T, x_i) = g(x_i).$$

- For each (t_k, x_i) the problem defined by $v_{\Delta t, \Delta x}(t_k, x_i)$ can have several solutions. In order to get uniqueness, for $\varepsilon > 0$, consider the entropy penalized scheme

$$v_{\Delta t, \Delta x}^\varepsilon(t_k, x_i) = \inf_{p \in \mathcal{P}(\mathcal{G}_{\Delta x}(x_i))} \left\{ \sum_{x_j \in \mathcal{G}_{\Delta x}(x_i)} p_j \left[\Delta t \left[\frac{1}{2} \left| \frac{x_j - x_i}{\Delta t} \right|^2 + f(x_i) \right. \right. \right. \\ \left. \left. \left. + v_{\Delta t, \Delta x}^\varepsilon(t_{k+1}, x_j) + \varepsilon \log p_j \right] \right\},$$

$$v_{\Delta t, \Delta x}^\varepsilon(T, x_i) = g(x_i).$$

- ▶ For every t_k , x_i and $\varepsilon > 0$, the previous problem has a unique minimizer $p_{opt}(x_i, \cdot, t_k)$.

$p_{opt}(x_i, x_j, t_k)$ denotes “optimal probability” of moving from x_i to x_j at time t_k .

- ▶ If $(\Delta t_n, \Delta x_n, \varepsilon_n) \rightarrow 0$, $\Delta x_n / \Delta t_n \rightarrow 0$, and $\varepsilon_n |\log(\Delta x_n)| / \Delta t_n \rightarrow 0$, then for every compact set $K \subseteq \mathbb{R}^d$ we have

$$\sup_{(t, x) \in \mathcal{T}_n \times (\mathcal{G}_{\Delta x_n} \cap K)} \left| v_{\Delta t_n, \Delta x_n}^{\varepsilon_n}(t, x) - v(t, x) \right| \xrightarrow{n \rightarrow \infty} 0.$$

- ▶ The previous scheme and the convergence result can be extended to several interesting contexts.

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We consider the following discretization of the mean field game problem

$$v_{\Delta t, \Delta x}^\varepsilon(t_k, x_i) = \inf_{p \in \mathcal{P}(\mathcal{G}_{\Delta x}(x_i))} \sum_{x_j \in \mathcal{G}_{\Delta x}(x_i)} p_j [c_{i,j}(p_j, m_{\Delta t, \Delta x}^\varepsilon(t_k, \cdot)) + v_{\Delta t, \Delta x}^\varepsilon(t_{k+1}, x_j)]$$

$$v_{\Delta t, \Delta x}^\varepsilon(T, x_i) = g(x_i, m_{\Delta t, \Delta x}^\varepsilon(T, \cdot))$$

$$m_{\Delta t, \Delta x}^\varepsilon(t_{k+1}, x_j) = \sum_{x_i \in \mathcal{G}_{\Delta x}} p_{\text{opt}}(x_i, x_j, t_k) m_{\Delta t, \Delta x}^\varepsilon(t_k, x_i)$$

$$m_{\Delta t, \Delta x}^\varepsilon(0, x_i) = \tilde{m}_0(x_i) \quad \forall x_i \in \mathcal{G}_{\Delta x},$$

where \tilde{m}_0 is a discretization of the initial distribution,

$$c_{i,j}(p_j, m) = \Delta t \left[\frac{1}{2} \left| \frac{x_j - x_i}{\Delta t} \right|^2 + f(x_i, m) \right] + \varepsilon \log(p_j),$$

$p_{\text{opt}}(x_i, \cdot, t_k) \in \mathcal{P}(\mathcal{G}_{\Delta x}(x_i))$ is the unique minimizer of $v_{\Delta t, \Delta x}^\varepsilon(t_k, x_i)$.

- ▶ The previous system is a particular instance of discrete time, finite state space MFGs introduced in Gomes-Mohr-Souza'10.
- ▶ Existence of a solution $(v_{\Delta t, \Delta x}^\varepsilon, m_{\Delta t, \Delta x}^\varepsilon)$ follows from a fixed-point argument (see Gomes-Mohr-Souza'10).
- ▶ If f and g are monotone, i.e. for $h = f, g$,

$$\sum_{x_j \in \mathcal{G}_{\Delta x}} [h(x_j, m_1) - h(x_j, m_2)] (m_1 - m_2) \geq 0 \quad \forall m_1, m_2 \in \mathcal{P}(\mathcal{G}_{\Delta x}),$$

it is possible to show that the solution is unique.

- ▶ Under this monotonicity assumption we will show later how to compute the equilibrium $(v_{\Delta t, \Delta x}^\varepsilon, m_{\Delta t, \Delta x}^\varepsilon)$ of (MFG_f) .

- ▶ Consider a sequence $(\Delta t_n, \Delta x_n, \varepsilon_n) \rightarrow 0$, set

$$v^n = v_{\Delta t_n, \Delta x_n}^{\varepsilon_n}, \quad m^n = m_{\Delta t_n, \Delta x_n}^{\varepsilon_n},$$

and p_{opt}^n the optimal transition probabilities.

- ▶ Set $m^{*,n}$ for the probability measure on Γ induced by \tilde{m}_0^n and p_{opt}^n .

The main result here is the following.

Theorem: Suppose that $\Delta x_n / \Delta t_n \rightarrow 0$, and $\varepsilon_n |\log(\Delta x_n)| / \Delta t_n \rightarrow 0$. Then the following holds

(i) Assume the “weak assumption” on the data f , g and m_0 . Then, any limit point $m^* \in \mathcal{P}(\Gamma)$ of $(m^{*,n})_{n \in \mathbb{N}}$ is a mean field game equilibrium, i.e. it satisfies (*).

(ii) Assume the “strong assumption” on the data f , g and m_0 . Then, if $m^{*,n} \rightarrow m^*$, we have that $(v^n, m^n) \rightarrow (v, \rho)$, where (v, ρ) is the solution to (MFG) associated to m^* .

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We consider the following “fictitious play” procedure to solve the discrete problem.

- ▶ Consider an arbitrary initial sequence of time marginals

$$M^1 = (M_0^1, \dots, M_N^1) \text{ and let } \bar{M}^1 = M^1.$$

- ▶ For $\ell \geq 1$ compute

$$V_k^\ell = \text{HJB}(V_{k+1}^\ell, \bar{M}_k^\ell), \quad V_N^\ell = g(\bar{M}_N^\ell)$$

$$\text{and then } M_{k+1}^{\ell+1} = \text{EV}(M_k^\ell, V_{k+1}^\ell), \quad M_0^{\ell+1} = m_0.$$

Set

$$\bar{M}^{\ell+1} := \frac{1}{\ell+1} \sum_{\ell'=1}^{\ell} M^{\ell'}.$$

- ▶ In terms of the best response (BR), the method can be written as

$$M^{\ell+1} = \text{BR}(\bar{M}^\ell).$$

Theorem: [Hadikhanloo-S'19] If f and g are monotone and Lipschitz w.r.t. to the second argument, then $(V^\ell, M^\ell, \bar{M}^\ell) \rightarrow (v^n, m^n, \bar{m}^n)$.

Example:

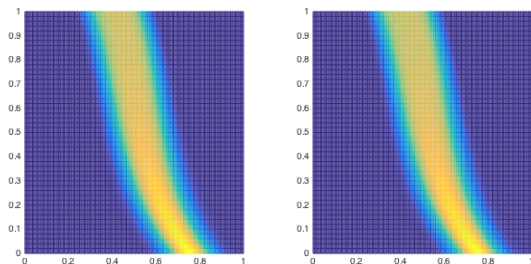
- ▶ Set $d = 1$, $T = 1$, $\rho_\sigma(z) = e^{-z^2/2\sigma^2} / \sqrt{2\pi\sigma^2}$, with $\sigma = 0.25$, and

$$f(x, m) = 2(x - 0.5)^2 + (\rho_\sigma * m) * \rho_\sigma(x)$$

$$g(x, m) = 2(x - 0.2)^2 + (\rho_\sigma * m) * \rho_\sigma(x)$$

$$m_0(x) = \frac{h(x)}{\int_0^1 h(x') dx'} \mathbb{I}_{[0,1]}(x), \quad \text{with } h(x) := e^{-\frac{(x-0.75)^2}{0.02}}$$

- ▶ Discretization parameters: $\Delta x = 0.005$, $\Delta t = 0.02$ and $\varepsilon = 0.002$.
- ▶ We apply the fictitious play procedure to (MFG^f) .



\bar{M}^ℓ (left) versus its best response $M^{\ell+1}$ (right), at step $\ell = 1000$.

- ▶ We have also tested the intuitive procedure $M^{\ell+1} = BR(M^\ell)$. Convergence fails in general. Indeed, there are configurations M such that $M = BR(BR(M))$ and $M \neq BR(M)$.

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Extensions (ongoing work with J. Gianatti)

- ▶ The previous analysis can be adapted to dynamics having the form

$$\dot{\gamma}(t) = A(\gamma(t)) + B(\gamma(t))\alpha(t).$$

For particular instances of the previous dynamics, the existence of MFG equilibria has been addressed in Cannarsa-Mendico'19 and Achdou-Mannucci-Marchi-Tchou'19.

- ▶ State constraints (Cannarsa-Capuani'18)

$$\gamma(t) \in K \quad \forall t \in [0, T].$$