

Numerical approaches for mean field games problems

Gabriel Turinici

Université Paris Dauphine

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Outline

- 1 Mathematical framework
 - Mean field games setting
 - Available literature
- 2 General monotonic algorithms (J. Salomon, G.T.)
 - Related applications: bi-linear problems
 - Framework
 - Construction of monotonic algorithms
- 3 Technology choice modelling (A. Lachapelle, J. Salomon, G.T.)
 - The model
 - Numerical simulations

Mean field games

- Mean field games: limits of Nash equilibriums for infinite number of players (P.L.Lions & J.M.Lasry)
- equation for each player $dX_t = \alpha dt + \sigma dW_t$, $\alpha(t, x) = \text{control}$
- $m(t, x) =$ the density of players at time t and position $x \in Q$
- evolution equation

$$\frac{\partial}{\partial t} m(t, x) - \nu \Delta m(t, x) + \text{div}(\alpha(t, x) m(t, x)) = 0,$$

$$m(0, x) = m_0(x).$$

- We consider the **optimisation setting**: $\min_{\alpha} J(\alpha)$

$$J(\alpha) := \Psi(m(\cdot, T)) + \int_0^T \left\{ \Phi(m(t, \cdot)) + \int_Q L(x, \alpha) m(t, x) dx \right\} dt$$

- Φ, Ψ can be linear, concave, ... Typical $L : L(x, \alpha) = \frac{\alpha^2}{2}$.

Numerics of MFG : literature overview

- (in)finite horizon: finite-difference discretization: approximation properties, existence and uniqueness, bounds on the solutions.
"Mean Field Games: Numerical Methods" Y. Achdou & I. Capuzzo-Dolcetta
<http://hal.archives-ouvertes.fr/hal-00392074/en/>
- Y. Achdou & I. Capuzzo-Dolcetta: Newton method for the coupled direct-adjoint critical point equations (finite horizon)
- O. Gueant: study of a prototypical case: solution, stability (in print in JMPA: "A reference case of mean field games"; available on www)
- solution of the MFG equations from an optimization point of view (A. Lachapelle, J. Salomon, G. Turinici, M3AS 2010)

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Optimal control of a Fokker-Plank equation (G. Carlier & J. Salomon)

Evolution equation :

$$\partial_t \rho - \epsilon^2 \Delta \rho + \operatorname{div}(v \rho) = 0 \quad (1)$$

$$\rho(x, t = 0) = \rho_0(x) \quad (2)$$

- goal: minimize w.r. to v the functional (for some given $V(\cdot)$) :

$$E(v) = \int \int \rho v^2 dx dt + \int \rho(x, 1) V(x)$$

Time dependent Schrödinger equation w. **BILINEAR** interaction (e.g. laser)

$$\begin{cases} i \frac{\partial}{\partial t} \Psi(x, t) = (H_0 - \epsilon(t)^k \mu(x)) \Psi(x, t) \\ \Psi(x, t = 0) = \Psi_0(x) \end{cases} \quad (3)$$

- vectorial case (rotation control, NMR):

$$i \frac{\partial}{\partial t} \Psi(x, t) = [H_0 + (E_1(t)^2 + E_2(t)^2) \mu_1 + E_1(t)^2 \cdot E_2(t) \mu_2] \Psi(x, t).$$

$H_0 = -\Delta + V(x)$, unbounded domain

Evolution on the unit sphere: $\|\Psi(t)\|_{L^2} = 1, \forall t \geq 0$.

- evaluation of the quality of a control through a objective functional to minimize

$$J(\epsilon) = -2\Re \langle \psi_{target} | \psi(\cdot, T) \rangle + \int_0^T \alpha(t) \epsilon^2(t) dt$$

$$J(\epsilon) = \|\psi_{target} - \psi(\cdot, T)\|_{L^2}^2 - 2 + \int_0^T \alpha(t) \epsilon^2(t) dt$$

$$J(\epsilon) = -\langle \Psi(T), O\Psi(T) \rangle + \int_0^T \alpha(t) \epsilon^2(t) dt$$

General monotonic algorithms (J. Salomon, G.T.)

state $X \in H$, control $v \in E$, $H, E =$ Banach spaces.

- $\partial_t X + A(t, v(t))X = B(t, v(t))$
- $\min_v J(v)$, $J(v) := \int_0^T F(t, v(t), X(t)) dt + G(X(T))$.
- $F, G: C^1 +$ **concavity** with respect to X (not v !)

$$\forall X, X' \in H, G(X') - G(X) \leq \langle \nabla_X G(X), X' - X \rangle$$

$\forall t \in \mathbb{R}, \forall v \in E, \forall X, X' \in H :$

$$F(t, v, X') - F(t, v, X) \leq \langle \nabla_X F(t, v, X), X' - X \rangle.$$

Direct-adjoint equations and first lemma

$$\partial_t X + A(t, v(t))X = B(t, v(t))$$

$$X(0) = X_0$$

$$\partial_t Y_v - A^*(t, v(t))Y_v + \nabla_X F(t, v(t), X_v(t)) = 0$$

$$Y_v(T) = \nabla_X G(X_v(T)).$$

Lemma

Suppose that A, B, F are differentiable everywhere in $v \in E$, then there exists $\Delta(\cdot, \cdot; t, X, Y) \in C^0(E^2, E)$ such that, for all $v, v' \in E$

$$J(v') - J(v) \leq \int_0^T \Delta(v', v; t, X_{v'}, Y_v) \cdot_E (v' - v) dt \quad (4)$$

Well-posedness

$$J(v') - J(v) \leq \int_0^T \Delta(v', v; t, X_{v'}, Y_v) \cdot_E (v' - v) dt \quad (5)$$

Remark: useful factorisation because can test at each step if J goes the right way; also can choose $v'(t^*) = v(t^*)$ if pb.

Remark: $\Delta(v', v; t, X, Y)$ has an explicit formula once the problem is given; also note the dependence on Y_v any not $Y_{v'}$.

Lemma

Under hypothesis on $A, B, F, G, \theta > 0$

$$\Delta(v', v; t, X, Y) = -\theta(v' - v) \quad (6)$$

has an unique solution $v' = \mathcal{V}_\theta(t, v, X, Y) \in E$.

Well-posedness

Theorem

Under hypothesis ...

- *the following eq. has a solution:*

$$\partial_t X_{v'}(t) + A(t, v')X_{v'}(t) = B(t, v') \quad (7)$$

$$v'(t) = \mathcal{V}_\theta(t, v(t), X_{v'}(t), Y_v(t)) \quad (8)$$

$$X_{v'}(0) = X_0 \quad (9)$$

- $\exists (\theta_k)_{k \in \mathbb{N}}$ such that $v^{k+1}(t) = \mathcal{V}_{\theta_k}(t, v^k(t), X_{v^{k+1}}(t), Y_{v^k}(t))$
- $J(v^{k+1}) - J(v^k) \leq -\theta_k \|v^{k+1} - v^k\|_{L^2([0, T])}^2$;
- if $v^{k+1}(t) = v^k(t) : \nabla_v J(v^k) = 0$.

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The Model : framework

- large economy: **continuum** of consumer agents
- time period: $[0, T]$
- any household owns exactly one house and cannot move to another one until T

The Model : the agents

- **arbitrage** between insulation and heating. A generic player (agent) has an insulation level $x \in [0, 1]$ ($x = 0$: no insulation, $x = 1$: maximal insulation)
- controlled process of the agent: $dX_t = \sigma dW_t + v_t dt + dN_t(X_t)$ where v is the **control** parameter (insulation effort), the noise level σ is given.
- note that X_t is a diffusion process with reflexion, in the above equality, $dN_t(X_t)$ has the form $\chi_{\{0,1\}}(X_t) \vec{n} d\xi_t$ (ξ = local time at the boundary $\{0, 1\} = \partial[0, 1]$ cf. Freidlin)
- initial density: $X_0 \sim m_0(dx)$

The Model : the costs

An agent of the economy solves a minimization problem composed of several terms:

- *Insulation acquisition cost*: $h(v) := \frac{v^2}{2}$
- *Insulation maintenance cost*: $g(t, x, m) := \frac{c_0 x}{c_1 + c_2 m(t, x)}$ increasing in x decreasing in m : **economy of scale, positive externality**. The agents should do the same choice, stay together. The higher is the number of players having chosen an insulation level, the lower are the related costs.
- *Heating cost*: $f(t, x) := p(t)(1 - 0,8x)$ where $p(t)$ is the unit heating cost (unit price of energy, say)

The model - The minimization problem and MFG (1)

- Define the aggregate state cost:

$$\Phi(m) := \int_0^1 \left(p(t)(1 - 0,8x) + \frac{c_0 x}{c_1 + c_2 m(t, x)} \right) m(t, x) dx$$

and $V = \Phi'$.

- In the model, the agents have **rational expectations**, i.e they see m as given; we can write the individual agent's problem:

$$\begin{cases} \inf_{v \text{ adm}} \mathbb{E} \left[\int_0^T h(v(t, X_t^x)) + V[m](X_t^x) dt \right] \\ dX_t = v_t dt + \sigma dW_t + dN_t(X_t), X_0 = x \end{cases}$$

The model - The minimization problem and MFG (2)

- We already know that it is linked with the optimal control problem:

$$\left\{ \begin{array}{l} \inf_{v \text{ adm}} \int_0^T \int_0^1 h(v(t, x)) + \Phi(m_t)(t) dt \\ \partial_t m - \frac{\sigma^2}{2} \Delta m + \operatorname{div}(vm) = 0, \quad m|_{t=0} = m_0(\cdot), \\ m'(\cdot, 0) = m'(\cdot, 1) = 0 \end{array} \right.$$

- Finally, if $\nu := \frac{\sigma^2}{2}$, a **Mean field equilibrium** (Nash equilibrium with an infinite number of players) corresponds to a solution of the following system:

$$\left\{ \begin{array}{l} \partial_t m - \nu \Delta m + \operatorname{div}(vm) = 0, \quad m|_{t=0} = m_0 \\ \nabla u = v \\ \partial_t u + \nu \Delta u + v \cdot \nabla u - \frac{u^2}{2} = \Phi'(m), \quad v|_{t=T} = 0 \end{array} \right. \quad (10)$$

The model - externality & scale effect

The MFG framework is interesting to describe a situation which lives between two economical ideas: **positive externality** and **economy of scale**

- **positive externality**: positive impact on any agent utility NOT INVOLVED in a choice of an insulation level by a player
- **economy of scale**: economies of scale are the cost advantages that a firm obtains due to expansion (unit costs decrease)

Criticism of the model:

- **stylised** from the "industrial" point of view
- not realistic (heating price, maintenance...)
- **transition effect** (continuous time, continuous space)
- **atomised** agent (her/his action has no influence on the global density, micro-macro approach)
- non-cooperative equilibrium with rational expectations

Numerical simulations

- Optimization method: **Monotonic algorithm**

$$\begin{cases} \partial_t m^{k+1} - \nu \Delta m^{k+1} + \operatorname{div}(v^{k+1} m^{k+1}) = 0, & m^{k+1}(x, 0) = m_0 \\ v^{k+1} = \frac{(\theta-1/2)v^k - \nabla u^k}{(\theta+1/2)} \\ \partial_t u^{k+1} + \nu \Delta u^{k+1} + v^{k+1} \cdot \nabla u^{k+1} - \frac{(u^{k+1})^2}{2} = \Phi'(m^{k+1}), & v^{k+1}(T) = \end{cases} \quad (11)$$

- Discretization of the PDEs: **Godunov scheme** (to preserve the positivity of the density m)

- The costs:

heating: $f(t, x) = p(t)(1 - 0,8x)$

insulation: $g(t, x, m) = \frac{x}{0.1+m(t,x)}$

- *1st example*: $p(t)$ constant / same choices
- *2d example*: $p(t)$ reaching a peak (non constant) / multiplicity of equilibria

Numerical results - First case

- the initial density of the householders is a gaussian centered in $\frac{1}{2}$
- the time period and the noise are respectively $T = 1$ and $\nu = 0.07$
- the energy price is constant ($p(t) \equiv 0, 3.2$ and 10)

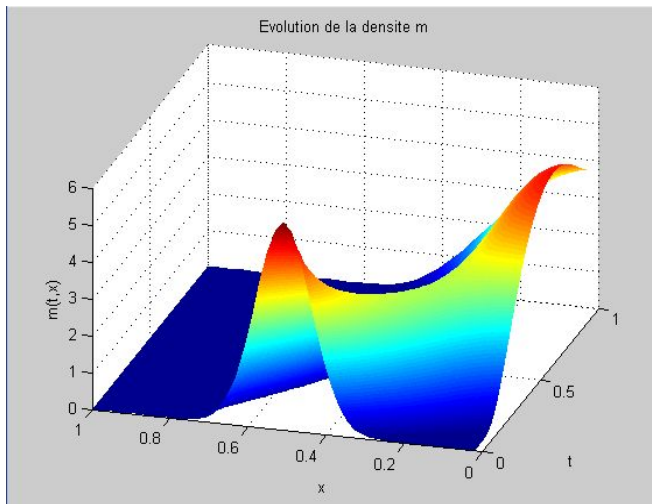


Figure: Numerical results : $p(t) \equiv 0$. Since the cost of energy is null all agents choose to heat their house, move to this choice together.

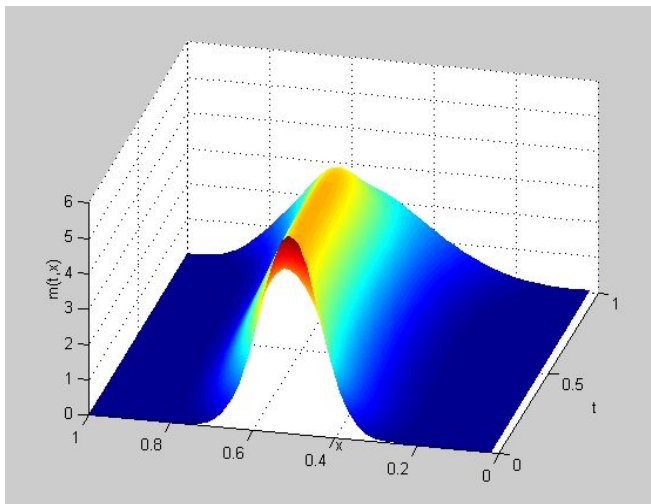


Figure: Numerical results : $p(t) \equiv 3.2$. Cost of energy is intermediary, agents keep their status.

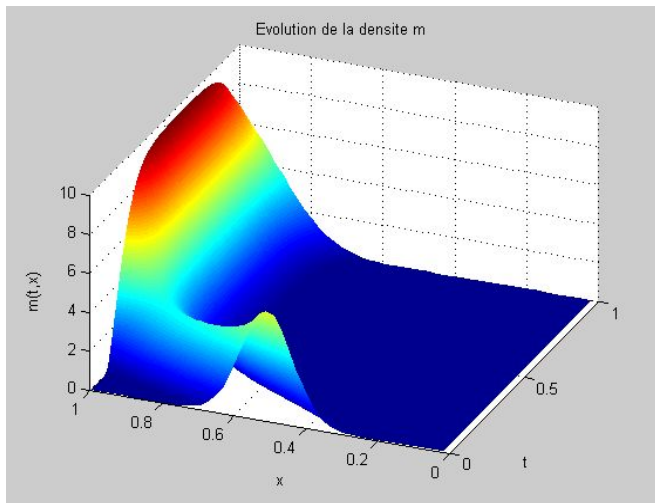


Figure: Numerical results : $p(t) \equiv 10$. Cost of energy is high, agents choose to better insulate, all have the same behavior.

Numerical results - Second case

- the initial density of the agents is an approximation of a Dirac in 0.1 (*i.e* agents are not equipped in insulation material)
- the energy price is **not a constant parameter**, we look at the following case: the price first **reaches a peak** and then decreases to its initial level.

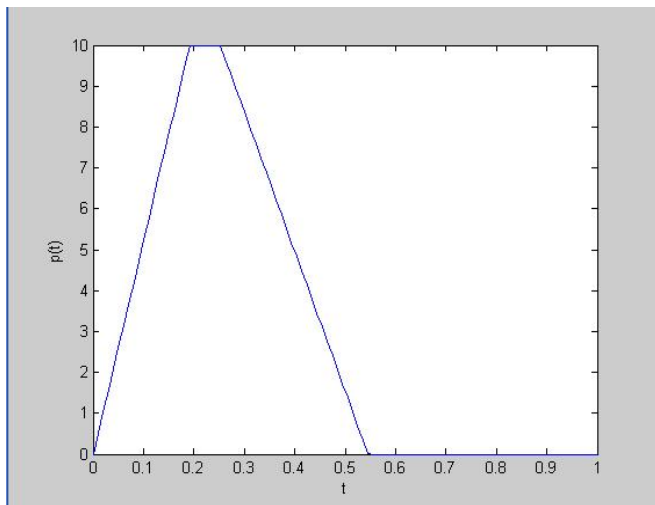


Figure: Numerical results - $p(t)$. Question: In such a case, can we find two Mean Field equilibria, the first related to the expectation of a higher insulation level, the second to the expectation of heating ?

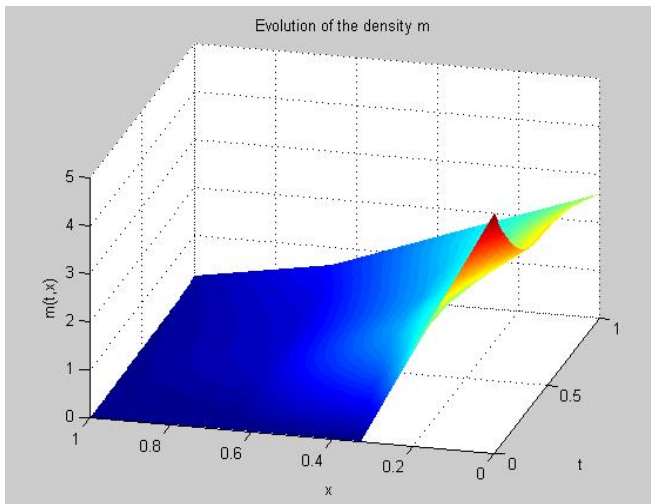


Figure: Numerical results - One of the two equilibria: the energy consumption equilibrium. Agents expect that everybody will keep a low insulation level so there are no gains in insulating.

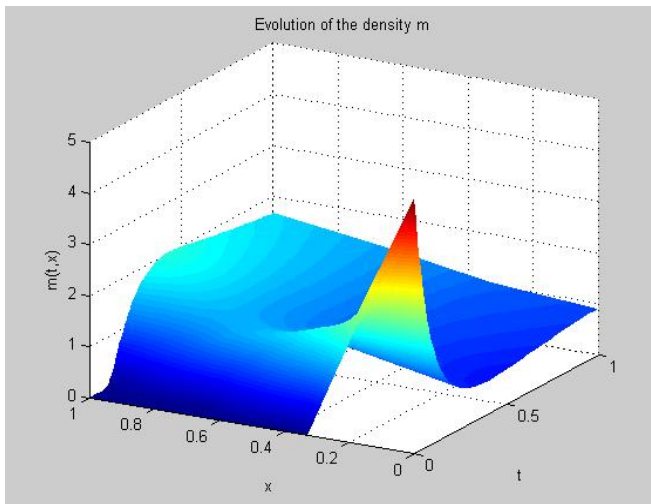


Figure: Numerical results - One of the two equilibria: the insulation equilibrium. Agents expect that everybody will better insulate, which makes insulating attractive.

Multiplicity of equilibria - Incentive policy

- we found an **insulation-equilibrium** and an **energy consumption-equilibrium**
- from the ecological point of view: the best is the insulation-equilibrium
- **incentive public policies** could steer towards the "best" equilibrium (from a certain point of view) when the solution is not unique.