

Newton schemes for mean field games

E. Carlini
Sapienza Università di Roma

joint work with
F.J. Silva, A.Zorkot

Valparaíso, Chile

Numerical methods for optimal transport problems,
mean field games, and multi-agent dynamics
January 8-12, 2024

Outline

- 1 Mean Field Games (MFG)
- 2 Newton-Semi Lagrangian scheme
- 3 Newton-Finite Difference scheme
- 4 Numerical Simulations

Table of Contents

- 1 Mean Field Games (MFG)
- 2 Newton-Semi Lagrangian scheme
- 3 Newton-Finite Difference scheme
- 4 Numerical Simulations

Mean Field Game Problem

Model introduced by Lasry, Lions and Huang, Malhamé, Caines, 2006

$$\begin{cases} -\partial_t u - \frac{\sigma^2}{2} \Delta u + H(x, \nabla u) = F(x, m(t)) & \text{in } [0, T) \times \mathbb{T}^d, \\ \partial_t m - \frac{\sigma^2}{2} \Delta m - \operatorname{div}(\partial_p H(x, \nabla u) m) = 0 & \text{in } (0, T] \times \mathbb{T}^d, \\ u(T, \cdot) = u_T(x), \quad m(\cdot, 0) = m_0 & \text{in } \mathbb{T}^d. \end{cases} \quad \text{(MFG)}$$

- $T > 0$, $\sigma \in \mathbb{R} \setminus \{0\}$, $H(x, p)$ is convex and differentiable w.r.t. p
- in the 1st line we have a Hamilton Jacobi Bellman (HJB) equation backward in time
- in the 2nd line we have a Fokker Plank equation forward in time
- m_0 and then $m(t, x)$ represents the density of a probability measure

Mean Field Game Problem

Model introduced by Lasry, Lions and Huang, Malhamé, Caines, 2006

$$\begin{cases} -\partial_t u - \frac{\sigma^2}{2} \Delta u + H(x, \nabla u) = F(m) & \text{in } \mathbb{T}^d \times [0, T), \\ \partial_t m - \frac{\sigma^2}{2} \Delta m - \operatorname{div}(\partial_p H(x, \nabla u) m) = 0 & \text{in } \mathbb{T}^d \times (0, T], \\ u(T, \cdot) = u_T(x), \quad m(0, \cdot) = m_0 & \text{in } \mathbb{T}^d. \end{cases} \quad (\text{MFG})$$

Our aim is

- **to propose a new numerical scheme by discretizing a Newton method in infinite dimension**

Some references of Numerical Approximation of MFG

- Y.Achdou, I.Capuzzo-Dolcetta ('10), Y.Achdou, F.Camilli, I.Capuzzo-Dolcetta ('12), Semi-implicite Finite Difference scheme, Newton Iteration
- E.C., F.J.Silva ('14, '15) Semi-Lagrangian scheme
- Y.Achdou, M.Lauriere ('20), Combine Continuation Methods with Newton Iterations
- H. Li, Y. Fan, and L. Ying ('21). Multiscale method for mean field games. Second order accurate
- F. Camilli, Q. Tang ('23) **Newton's method in infinite dimension for Mean Field Games**

Assumptions

We assume $\sigma \neq 0$ and for $\alpha \in (0, 1)$

(H1) m_0 is n positive, $m_0 \in \mathcal{P}(\mathbb{T}^d) \cap C^{2+\alpha}(\mathbb{T}^d)$, $u_T \in C^{2+\alpha}(\mathbb{T}^d)$

(H2) F, F', F'' are uniformly bounded mappings from $\mathbb{R}^+ \rightarrow \mathbb{R}$.

Moreover, $F'(\cdot) \geq 0$

(H3) $H : \mathbb{T}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$ continuous, two times differentiable in p and there exists $c, C > 0$ such that

$$cI \leq H_{pp}(x, p) \leq CI, \quad (x, p) \in \mathbb{T}^d \times \mathbb{R}^d$$

Under **(H1)-(H3)** the MFG system admits one **classical solution**

Newton method

Following (Camilli Tang 2023) we define the map

$$\mathcal{T} : (u, m) \rightarrow \begin{pmatrix} -\partial_t u - \frac{\sigma^2}{2} \Delta u + H(Du) - F(m) \\ \partial_t m - \frac{\sigma^2}{2} \Delta m - \operatorname{div}(m H_p(Du)) \\ u(T) - u_T(x) \\ m(0) - m_0(x) \end{pmatrix}.$$

Then system (**MFG**) is equivalent to

$$\mathcal{T}(u, m) = 0$$

and the corresponding **Newton's iterations** can be written, at a generic iteration n , as

$$J\mathcal{T}(u^{n-1}, m^{n-1})(u^n - u^{n-1}, m^n - m^{n-1}) = -\mathcal{T}(u^{n-1}, m^{n-1}). \quad (1)$$

Newton method

The Newton method reads:

given (u^0, m^0) , find (u^n, m^n) by solving (2) for $n \geq 1$

Theorem (Camilli Tang 2023)

If the initial guess (u^0, m^0) is close enough to the (u, m) solution of (MFG), then

$$\|u - u^n\|_{C^{0,1}} + \|m - m^n\|_{C^0} \leq C^* (\|u - u^{n-1}\|_{C^{0,1}} + \|m - m^{n-1}\|_{C^0})^2. \quad (3)$$

Notation:

$$\|u\|_{C^{0,1}} = \|u\|_{C^0} + \|Du\|_{C^0}$$

For simplicity, next we consider the the Eikonal HJB equation

$$H(x, p) = \frac{p^2}{2} - V(x)$$

Table of Contents

- 1 Mean Field Games (MFG)
- 2 Newton-Semi Lagrangian scheme
- 3 Newton-Finite Difference scheme
- 4 Numerical Simulations

Linearized HJ

Given

$$L(t, x) = \frac{|q^n(t, x)|^2}{2} + F(m^{n-1}(t, x)) + F'(m^{n-1}(t, x))(m^n(t, x) - m^{n-1}(t, x))$$

let us consider

$$\begin{cases} -\partial_t u^n - \frac{\sigma^2}{2} \Delta u^n + q^n Du^n - L(t, x) = 0 \\ u^n(T, x) = u_T(x) \end{cases} \quad (4)$$

Representation formula for $u^n(t, x)$ holds true

$$u^n(t, x) = \mathbb{E} \left(\int_t^T L(s, X^{t,x}(s)) ds + u_T(X^{t,x}(T)) \right),$$

where

$$X^{t,x}(s) := x - \int_t^s q^n(r, X^{t,x}(r)) dr + \int_t^s \sigma dW(r) \quad \text{for all } s \in [t, T].$$

Main Ingredients for SL scheme

- Representation formula in $[t_k, t_{k+1}]$

$$u^n(t_k, x) = \mathbb{E} \left(\int_{t_k}^{t_{k+1}} L(s, X^{t_k, x}(s)) ds \right) + u^n(t_{k+1}, X^{t_k, x}(t_{k+1}))$$

- Semi discretization in time by one-step weak Euler:
Let us approximate the stochastic characteristics by

$$X^{t_k, x}(t_{k+1}) \simeq x - \Delta t q^n(t_k, x) + \sigma \Delta W,$$

where $P(\Delta W = \pm \sqrt{\Delta t}) = \frac{1}{2}$

- Trapezoidal rule for Running cost

$$\int_{t_k}^{t_{k+1}} L(s, X^{t_k, x}(s)) ds \simeq \Delta t L(t_k, x)$$

Ref. Camilli, Falcone(1998), Falcone, Ferretti (2014)

Fully discrete SL scheme for HJL

Let us define $\{u_{k,i}^n\}$ as the solution to

$$\begin{cases} u_{k,i}^n = \frac{1}{2} \sum_{\pm} I[u_{k+1}^n](x_i - \Delta t q^n(t_k, x_i) \pm \sqrt{\Delta t}) + \Delta t L(t_k, x_i) \\ u_{N\Delta t, i}^n = u_T(x_i). \end{cases} \quad (\text{SL})$$

where for a given grid function f

$$I[f](x) := \sum_j \beta_j(x) f_j$$

where $\{\beta_j\}$ denotes the \mathbb{P}_1 basis of **piecewise linear function**, and

$$L(t_k, x_i) = \frac{|q^n(t_k, x_i)|^2}{2} + V(x_i) + F(m_{k,i}^{n-1}) + F'(m_{k,i}^{n-1})(m_{k,i}^n - m_{k,i}^{n-1})$$

where $m_{k,i}^n \simeq m^n(x_i, t_k)$

Backward explicit SL scheme for HJL

For given $Q_k^n = \{q^n(t_k, x_i)\}$, $M_k^n = \{m_{k,i}^n\}$, compute $U_k^n = \{u_{k,i}^n\}$

$$\begin{cases} U_k^n = \mathcal{A}(Q_k^n)U_{k+1}^n + \Delta t \mathcal{W}_k M_k^n + \Delta t \mathcal{B}_k & k = 0, \dots, N_{\Delta t} - 1, \\ U_{N_{\Delta t}}^n = U_T \end{cases} \quad (5)$$

where, given $\{m_{k,i}^{n-1}\}$, we define

$$\begin{cases} (\mathcal{A}(Q))_{ij} := \frac{1}{2} \sum_{\pm} \beta_j(x_i - \Delta t(Q)_i \pm \sqrt{\Delta t}) \\ (\mathcal{W}_k)_{i,j} := F'(m_{k,i}^{n-1})\delta_{i,j} \\ (\mathcal{B}_k)_i := \frac{|q^n(t_k, x_i)|^2}{2} + V(x_i) + F(m_{k,i}^{n-1}) + F'(m_{k,i}^{n-1})m_{k,i}^{n-1} \end{cases}$$

Fully discrete adjoint SL scheme for FP

Given

$$G(t, x) = \operatorname{div}(m^{n-1}(t, x)(Du^n(t, x) - Du^{n-1}(t, x)))$$

let us consider

$$\begin{cases} \partial_t m^n - \frac{\sigma^2}{2} \Delta m^n - \operatorname{div}(m^n q^n) = G(t, x) \\ m^n(x, 0) = m_0(x). \end{cases}$$

Using the **duality** property

$$\int L(f)g dx = \int L^*(g)f dx$$

of the operators

$$L^*(m) := -\frac{\sigma^2}{2} \Delta m - \operatorname{div}(q(x)m)$$

$$L(u) := -\frac{\sigma^2}{2} \Delta u + q(x)^\top Du$$

we derive a scheme for the FP

Adjoint SL scheme for FP

For given $Q_k^n = \{q^n(t_k, x_i)\}$, $U_k^n = \{u_{k,i}^n\}$, compute $M_k^n = \{m_{k,i}^n\}$

$$\begin{cases} M_{k+1}^n = \mathcal{A}^*(Q_k^n)M_k^n + \Delta t \mathcal{Z}_k U_k^n + \Delta t \mathcal{C}_k, & k = 0, \dots, N_{\Delta t} - 1 \\ M_0^n = M_0, \end{cases}$$

where $\mathcal{A}^*(Q)$ is the transpose of $\mathcal{A}(Q)$ and, for given M_k^{n-1} , \mathcal{Z}_k is the matrix defined such that

$$\begin{aligned} (\mathcal{Z}_k U)_i &= (D_{\Delta x} M_k^{n-1})_i (D_{\Delta x} U_k)_i + (M_k^{n-1})_i (\text{Lap}_{\Delta x} U_k)_i \\ &\simeq \text{div}(m^{n-1}(t_k, x_i) (Du^n(t_k, x_i))) \end{aligned}$$

and

$$\begin{aligned} (\mathcal{C}_k)_i &= -(D_{\Delta x} M_k^{n-1})_i (Q_k^n)_i - (M_k^{n-1})_i (\text{div}_{\Delta x} Q_k^n)_i \\ &\simeq -\text{div}(m^{n-1}(t_k, x_i) q(t_k, x_i)) \end{aligned}$$

Fully discrete Newton scheme

Given (U^{n-1}, M^{n-1}) , define $Q_k^n := D_h U_k^{n-1}$ and compute (U^n, M^n) as solution of the linear system

$$\begin{cases} U_k^n = \mathcal{A}(Q_k^n)U_{k+1}^n + \Delta t \mathcal{W}_k M_k^n + \Delta t \mathcal{B}_k & k = 0, \dots, N_{\Delta t} - 1, \\ M_{k+1}^n = \mathcal{A}^*(Q_k^n)M_k^n + \Delta t \mathcal{Z}_k U_k^n + \Delta t \mathcal{C}_{k+1}, & k = 0, \dots, N_{\Delta t} - 1 \\ U_{N_{\Delta t}}^n = U_T \quad M_0^n = M_0, \end{cases} \quad (\text{Newton-SL})$$

which can be written as an Hamiltonian system

$$\begin{pmatrix} \mathbb{A} & -\mathbb{W} \\ -\mathbb{Z} & -\mathbb{A}^\top \end{pmatrix} \begin{pmatrix} \bar{U} \\ \bar{M} \end{pmatrix} = \begin{pmatrix} \mathbb{B} \\ \mathbb{C} \end{pmatrix} \quad (6)$$

Proposition

If $M^n \geq 0$, then for any $n \in \mathbb{N}$ there exists a unique solution (U^n, M^n) to (6)

Table of Contents

- 1 Mean Field Games (MFG)
- 2 Newton-Semi Lagrangian scheme
- 3 Newton-Finite Difference scheme
- 4 Numerical Simulations

Implicit FD scheme for HJL

Given q^n, m^n, m^{n-1} , we define $\{u_{k,i}^n\}$ for $k = 0, \dots, N_{\Delta t} - 1$ as the solution to the following Implicit FD scheme

$$\begin{cases} u_{k,i}^n = u_{k+1,i}^n + \Delta t \mu_i^k \text{Lapu}_{k,i}^n + \Delta t q^n(t_k, x_i) D_h u_{k,i}^n + \Delta t L(t_k, x_i) \\ u_{N_{\Delta t}, i}^n = u_T(x_i). \end{cases} \quad (\text{FD})$$

where

$$\mu_i^k = \frac{\sigma^2}{2} + \Delta x |q^n(t_k, x_i)|$$

and

$$L(t_k, x_i) = \frac{|q^n(t_k, x_i)|^2}{2} + V(x_i) + F(m_{k,i}^{n-1}) + F'(m_{k,i}^{n-1})(m_{k,i}^n - m_{k,i}^{n-1})$$

which can be rewritten as

$$\begin{cases} U_k^n = U_{k+1}^n - \Delta t \mathcal{D}_k U_k^n + \Delta t \mathcal{W}_k M_k^n + \Delta t \mathcal{B}_k & k = 0, \dots, N_{\Delta t} - 1, \\ U_{N_{\Delta t}}^n = U_T \end{cases}$$

FD Newton scheme

Given (U^{n-1}, M^{n-1}) , define $Q_k^n := D_h U_k^{n-1}$ and compute (U^n, M^n) as solution of the linear system, for $k = 0, \dots, N_{\Delta t} - 1$

$$\begin{cases} U_k^n = U_{k+1}^n - \Delta t(\mathcal{D}_k U_k^n + \mathcal{W}_k M_k^n + \mathcal{B}_k) \\ M_{k+1}^n = M_k^n - \Delta t((\mathcal{D}_k)^* M_{k+1}^n + \mathcal{Z}_{k+1} U_{k+1}^n + \mathcal{C}_{k+1}) \\ U_{N_{\Delta t}}^n = U_T \quad M_0^n = M_0, \end{cases}$$

(Newton-FD)

Proposition

If $M^n \geq 0$, then for any $n \in \mathbb{N}$ there exists a unique solution (U^n, M^n) to (Newton-FD)

Table of Contents

- 1 Mean Field Games (MFG)
- 2 Newton-Semi Lagrangian scheme
- 3 Newton-Finite Difference scheme
- 4 Numerical Simulations**

Newton SL Algorithm

Initial guesses M^0, U^0 and tolerances $\delta > 0, \tau > 0$

Compute $Q^0 = D_h u^0$, set $\epsilon_0 = 1, n = 0$

While $\epsilon_n > \tau$

 Compute M^n, U^n by (Newton-SL)
 (Gauss-Seidel with tolerance δ)

 Update $Q^n, n = n + 1$

 Let $\epsilon_n = \|M_n - M_{n+1}\|_\infty + \|U_n - U_{n+1}\|_\infty$

End While

Newton SL Algorithm

Initial guesses M^0, U^0 and tolerances $\delta > 0, \tau > 0$

Compute $Q^0 = D_h u^0$, set $\epsilon_0 = 1, n = 0$

While $\epsilon_n > \tau$

 Compute M^n, U^n by (Newton-SL)
 (Gauss-Seidel with tolerance δ)

 Update $Q^n, n = n + 1$

 Let $\epsilon_n = \|M_n - M_{n+1}\|_\infty + \|U_n - U_{n+1}\|_\infty$

End While

Newton FD Algorithm

Initial guesses M^0, U^0 and tolerances $\delta > 0, \tau > 0$

Compute $Q^0 = D_h u^0$, set $\epsilon_0 = 1, n = 0$

While $\epsilon_n > \tau$

 Compute M^n, U^n by (**Newton-FD**)
 (Gauss-Seidel with tolerance δ)

 Update $Q^n, n = n + 1$

 Let $\epsilon_n = \|M_n - M_{n+1}\|_\infty + \|U_n - U_{n+1}\|_\infty$

End While

Newton FD Algorithm

Initial guesses M^0, U^0 and tolerances $\delta > 0, \tau > 0$

Compute $Q^0 = D_h u^0$, set $\epsilon_0 = 1, n = 0$

While $\epsilon_n > \tau$

 Compute M^n, U^n by (**Newton-FD**)

 (Gauss-Seidel with tolerance δ)

 Update $Q^n, n = n + 1$

 Let $\epsilon_n = \|M_n - M_{n+1}\|_\infty + \|U_n - U_{n+1}\|_\infty$

End While

MFG 1d

Let us choose $H(p) = \frac{p^2}{2}$, $V(x) = 0$,

$$m_0(x) = \begin{cases} 4 \sin^2(2\pi(x - 1/4)) & \text{if } x \in [1/4, 3/4] \\ 0 & \text{otherwise} \end{cases}$$

$$F(m(x)) = 3m_0(x) - 4 \min(4, m), \quad u_T = 0.$$

We set $\sigma = 0.05$, $T = 0.05$, $\tau = \delta = 10^{-4}$ and periodic boundary conditions.

We solve the MFG system using

Newton-SL, **Newton-FD**, **FD-Newton** (Ref. Achodou, Capuzzo Dolcetta (2010))

and **Fixed Point-SL** (Carlini-Silva 2014)

Comparison Newton-SL vs Fixed Point-SL

h	$E_{\infty}(u)$	$E_{\infty}(m)$	Time	Iterations
Newton-SL				
$2.50 \cdot 10^{-2}$	$5.51 \cdot 10^{-2}$	$1.64 \cdot 10^{-1}$	0.61s	6
$1.25 \cdot 10^{-2}$	$2.40 \cdot 10^{-2}$	$1.16 \cdot 10^{-1}$	2.77s	7
$6.25 \cdot 10^{-3}$	$1.83 \cdot 10^{-2}$	$6.61 \cdot 10^{-2}$	13.92s	7
$3.125 \cdot 10^{-3}$	$4.50 \cdot 10^{-3}$	$1.41 \cdot 10^{-2}$	80.60s	7
Fixed Point-SL				
$2.50 \cdot 10^{-2}$	$5.75 \cdot 10^{-2}$	$1.62 \cdot 10^{-1}$	8.09s	10
$1.25 \cdot 10^{-2}$	$2.84 \cdot 10^{-2}$	$1.11 \cdot 10^{-1}$	40.79s	10
$6.25 \cdot 10^{-3}$	$2.15 \cdot 10^{-2}$	$5.84 \cdot 10^{-2}$	259.72s	12
$3.125 \cdot 10^{-3}$	$9.50 \cdot 10^{-3}$	$6.51 \cdot 10^{-3}$	2793.71s	12

Table: Errors for the approximation of solution (u, m) using Fixed Point and Newton SL schemes with $\Delta t = h^{3/2}/2$

Newton-FD vs FD-Newton

h	$E_\infty(u)$	$E_\infty(m)$	Time	Iterations
FD-Newton				
$2.50 \cdot 10^{-2}$	$1.23 \cdot 10^{-1}$	$3.11 \cdot 10^{-2}$	2.23s	7
$1.25 \cdot 10^{-2}$	$6.21 \cdot 10^{-2}$	$1.63 \cdot 10^{-2}$	18.32s	8
$6.25 \cdot 10^{-3}$	$3.14 \cdot 10^{-2}$	$8.75 \cdot 10^{-3}$	92.91s	8
$3.125 \cdot 10^{-3}$	$1.77 \cdot 10^{-2}$	$9.54 \cdot 10^{-3}$	597.21s	8
Newton-FD				
$2.50 \cdot 10^{-2}$	$1.532 \cdot 10^{-1}$	$3.42 \cdot 10^{-2}$	1.48s	7
$1.25 \cdot 10^{-2}$	$6.71 \cdot 10^{-2}$	$1.83 \cdot 10^{-2}$	12.27s	7
$6.25 \cdot 10^{-3}$	$3.37 \cdot 10^{-2}$	$9.51 \cdot 10^{-3}$	68.10s	7
$3.125 \cdot 10^{-3}$	$1.91 \cdot 10^{-2}$	$7.38 \cdot 10^{-3}$	436.01s	7

Table: Errors for the approximation of solution (u, m) using **FD-Newton** and **Newton-FD** schemes with $\Delta t = h/4$

MFG 1d

Let us choose $H(p) = p^2$, $V(x) = 200 \cos(2\pi x) - 10 \cos(4\pi x)$,

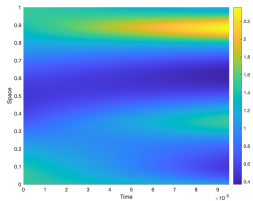
$$m_0(x) = 1 + \frac{1}{2} \cos(2\pi x), \quad u_T = \sin(4\pi x) + 0.1 \cos(10\pi x),$$

$$F(m(x)) = m^2, \quad .$$

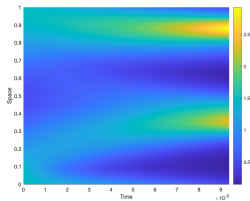
We set $T = 0.01$, $\tau = \delta = 10^{-4}$ and periodic boundary conditions.

We solve the MFG system using **Newton-SL**, **Newton-FD**, **FD-Newton** (Ref. Achodou. Capuzzo Dolcetta 2010) and We set $\Delta t = h$ for the two finite differences schemes, and $\Delta t = h^{3/2}/2$ for Newton-SL

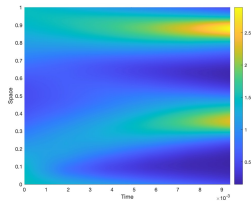
MFG 1d $\sigma = 0.4$



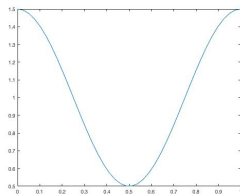
Newton-SL



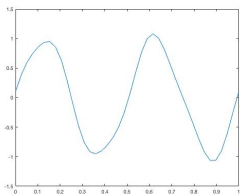
Newton-FD



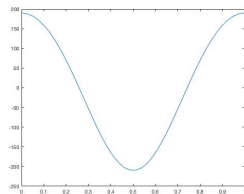
FD-Newton



Initial mass m_0

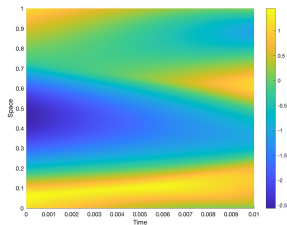


Terminal cost u_T

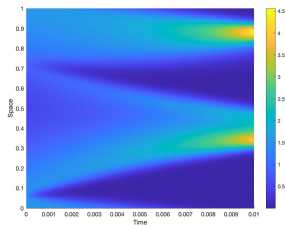


Potential V

MFG, Newton-SL $\sigma = 0.02$



Value Function



Density

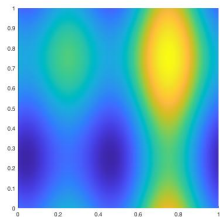
MFG 2d, Newton-SL

Let us choose $H(p) = p^2$, $V(x) = -\sin(2\pi x_1) - \sin(2\pi x_2) - \cos(4\pi x_1)$,
 $T = 1$, $\tau = \delta = 10^{-4}$, periodic boundary conditions,

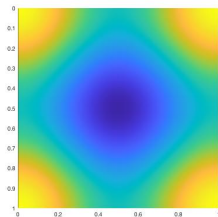
$$m_0(x) = 1 + 12 \cos(2\pi x_1) + 12 \cos(2\pi x_2), \quad u_T = \cos(2\pi x_1) + \cos(2\pi x_2),$$

$$F(m(x)) = m^2.$$

We solve the MFG system using **Newton-SL** with $\Delta t = h^2$

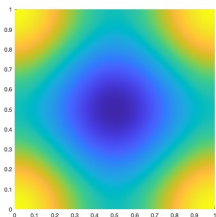


Potential V

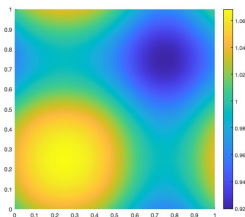


Terminal cost u_T

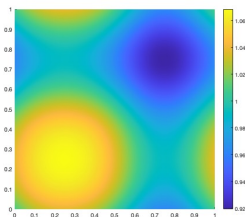
MFG 2d, Newton-SL



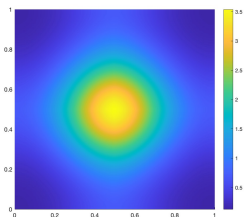
Initial condition m_0



Density at $t = 0.5$



Density at $t = 0.75$



Density at $T = 1$

Conclusions

- **FD-Newton** and **Newton-DF** show similar behaviour in terms of CPU time and accuracy
- **Newton-SL** needs the cheapest CPU time and shows comparable accuracy with respect to the other methods
- **Newton-SL** scheme works well in hyperbolic regime (σ small)

References

- F. Camilli and Q.Tang. *A convergence rate for the newton's method for mean field games with non-separable hamiltonians*, (2023)
- E.Carlini and F.J.Silva, *A Semi-Lagrangian scheme for a degenerate second order Mean Field Game system*, DCDS-A, (2015).
- Y. Achdou and I. Capuzzo-Dolcetta. Mean field games: Numerical methods. *SIAM Journal on Numerical Analysis*, 48, 01 (2010).
- E.Carlini, F.J.Silva and A. Zorkot *Newton methods for MFGs*, in preparation (2024).