

PDE Methods for Mean Field Games with Non-Separable Hamiltonian: Data in Sobolev Spaces (Continued)

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Steps of the energy method

- Introduce an approximate problem.
- Prove existence of solutions for the approximate problem.
- **Prove an estimate for these solutions which is uniform in the approximation parameter.**
- Use this estimate to help pass to the limit as the approximation parameter(s) vanish.
- Prove that the limiting solution solves the original problem.
- Make similar estimates for uniqueness.

The mean field games system

- For reference, we (again) write down the mean field games system we are considering:

$$u_t + \varepsilon H(t, x, m, Du) = -\Delta u, \quad (1)$$

$$m_t + \varepsilon \operatorname{div}(m H_p(t, x, m, Du)) = \Delta m, \quad (2)$$

in the context of the planning problem, i.e., with boundary conditions

$$m(0, \cdot) = m_0, \quad u(T, \cdot) = u_T. \quad (3)$$

The approximate problems

- We had set up the approximate problems with two approximating parameters, n and δ .
- We had let \mathbb{P}_δ be a mollifier in the x variable; a good choice is truncation of the Fourier series at the level $1/\delta$.
- We mollified the data:

$$\mu^{n,\delta}(0, \cdot) = \mathbb{P}_\delta \mu_0, \quad w^{n,\delta}(T, \cdot) = \mathbb{P}_\delta w_T.$$

- We had set up an iterative scheme:

$$w_t^{n+1,\delta} + \varepsilon P\Theta(t, x, \mu^{n,\delta}, Dw^{n,\delta}) + \Delta w^{n+1,\delta} = 0,$$

$$\mu_t^{n+1,\delta} + \varepsilon \operatorname{div} \left((\mu^{n,\delta} + \bar{m}) \Theta_p(t, x, \mu^{n,\delta}, Dw^{n,\delta}) \right) - \Delta \mu^{n+1,\delta} = 0.$$

- We initialized with

$$w^{0,\delta} = \mu^{0,\delta} = 0.$$

We skipped the uniform estimate

- The most important step in the energy method is finding an estimate, uniform in the approximation parameter(s).
- We will now establish that

$$\sup_{t \in [0, T]} \|\mu^{n, \delta}\|_{s-1}^2 + \sup_{t \in [0, T]} \|w^{n, \delta}\|_s^2 + \int_0^T \|\mu^{n, \delta}\|_s^2 + \|w^{n, \delta}\|_{s+1}^2 d\tau$$

is bounded by some constant (independent of n and δ).

- To do this, we will introduce a smallness constraint.

The theorem we are proving

- Remember the theorem I stated last time:

Theorem

Let $T > 0$ and $\varepsilon > 0$ be given. Let $s \geq \lceil \frac{d+5}{2} \rceil$ and let $\mu_0 \in H^{s-1}(\mathbb{T}^d)$ be such that $\bar{m} + \mu_0$ is a probability measure. Let $u_T \in H^s(\mathbb{T}^d)$ be given. Assume that the conditions **(H1)** and **(H2)** are satisfied. Then there exists $\mu \in L^\infty([0, T]; H^{s-1}) \cap L^2([0, T]; H^s)$ and there exists $u \in L^\infty([0, T]; H^s) \cap L^2([0, T]; H^{s+1})$ such that $\bar{m} + \mu$ is a probability measure for all $t \in [0, T]$, and such that $(u, \bar{m} + \mu)$ satisfies (1), (2), (3). Furthermore, for all $s' \in [0, s)$, we have $\mu \in C([0, T]; H^{s'-1})$ and $u \in C([0, T]; H^{s'})$.

The hypothesis **(H1)**

- **(H1)** The function \mathcal{H} is such that there exists a non-decreasing function $\tilde{F} : [0, \infty) \rightarrow [0, \infty)$ such that for all $\beta \in \mathbb{N}^{2d+1}$ with $|\beta| \leq s + 2$,

$$\left| \partial^\beta \Theta(\cdot, \cdot, \nu, Dy) \right|_\infty \leq \tilde{F} (|\nu|_\infty + |Dy|_\infty).$$

- This has the immediate consequence that Θ and its derivatives are bounded in terms of the norm of the solution:

$$\begin{aligned} & \sum_{|\beta| \leq 2} \left\| (\partial^\beta \Theta)(t, \cdot, \mu, Dw) \right\|_{s-1} \\ & \leq \bar{C} F \left(\|\mu\|_{\lceil \frac{d+1}{2} \rceil}^2 + \|Dw\|_{\lceil \frac{d+1}{2} \rceil}^2 \right) (1 + \|\mu\|_{s-1} + \|w\|_s)^{s-1}. \end{aligned}$$

Our norms to estimate

- For all $n \in \mathbb{N}$, we define M_n and N_n to be

$$M_n = \sup_{t \in [0, T]} (\|Dw^n\|_{s-1}^2) + \sup_{t \in [0, T]} (\|\mu^n\|_{s-1}^2),$$

$$N_n = \sum_{1 \leq |\alpha| \leq s} \int_0^T \|\partial^\alpha Dw^n\|_0^2 d\tau + \sum_{0 \leq |\alpha| \leq s-1} \int_0^T \|\partial^\alpha D\mu^n\|_0^2 d\tau.$$

- Our solutions actually depend on δ as well, but we will suppress this dependence for now.
- The N_n characterizes the parabolic gain of regularity; we will focus on estimating M_n and we will get an estimate for N_n as we go.

We begin with μ^n :

$$\begin{aligned}
 & \frac{d}{dt} \frac{1}{2} \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1})^2 dx \\
 = & \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) (\partial^\alpha \Delta \mu^{n+1}) dx - \varepsilon \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) \partial^\alpha (D\mu^n \cdot \Theta_p(\cdot, x, \mu^n, Dw^n)) dx \\
 & - \varepsilon \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) \partial^\alpha \left((\mu^n + \bar{m}) \sum_{i=1}^d \Theta_{x_i p_i}(\cdot, x, \mu^n, Dw^n) \right) dx \\
 & - \varepsilon \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) \partial^\alpha \left((\mu^n + \bar{m}) \sum_{i=1}^d [(\Theta_{qp_i}(\cdot, x, \mu^n, Dw^n)) (\partial_{x_i} \mu^n)] \right) dx \\
 - & \varepsilon \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) \partial^\alpha \left((\mu^n + \bar{m}) \sum_{i=1}^d \sum_{j=1}^d [(\Theta_{p_i p_j}(\cdot, x, \mu^n, Dw^n)) (\partial_{x_i x_j}^2 w^n)] \right) dx.
 \end{aligned}$$

We integrate with respect to time and rearrange:

$$\begin{aligned}
& \frac{1}{2} \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}(t, x))^2 dx - \frac{1}{2} \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}(0, x))^2 dx + \int_0^t \int_{\mathbb{T}^d} |D\partial^\alpha \mu^{n+1}|^2 dx d\tau \\
& \quad = -\varepsilon \int_0^t \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) \partial^\alpha (D\mu^n \cdot \Theta_p(\tau, x, \mu^n, Dw^n)) dx d\tau \\
& \quad \quad - \varepsilon \int_0^t \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) \partial^\alpha \left((\mu^n + \bar{m}) \sum_{i=1}^d \Theta_{x_i p_i}(\tau, x, \mu^n, Dw^n) \right) dx d\tau \\
& \quad \quad - \varepsilon \int_0^t \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) \partial^\alpha \left((\mu^n + \bar{m}) \sum_{i=1}^d [(\Theta_{q p_i}(\tau, x, \mu^n, Dw^n)) (\partial_{x_i} \mu^n)] \right) dx d\tau \\
& \quad \quad - \varepsilon \int_0^t \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) \partial^\alpha \left((\mu^n + \bar{m}) \sum_{i=1}^d \sum_{j=1}^d [(\Theta_{p_i p_j}(\tau, x, \mu^n, Dw^n)) (\partial_{x_i x_j}^2 w^n)] \right) dx d\tau \\
& \hspace{20em} = I + II + III + IV.
\end{aligned}$$

Work to do:

- We need to estimate these terms, I , II , III , and IV in terms of M_n and N_n .
- We will actually do some further decomposing: $I = I_A + I_B$, $III = III_A + III_B$, and $IV = IV_A + IV_B$.
- Each of these further decompositions involves separating out the leading term, as in an example in the last lecture.
- This is for μ ; we have corresponding work to do for w as well.

The term I_A

- We have the definition

$$I_A = -\varepsilon \int_0^t \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1})(\partial^\alpha D\mu^n) \cdot \Theta_p(\tau, x, \mu^n, Dw^n) \, dx d\tau$$

- We pull Θ_p through the integral:

$$I_A \leq \varepsilon |\Theta_p(t, \cdot, \mu^n, Dw^n)|_\infty \int_0^t \int_{\mathbb{T}^d} |\partial^\alpha \mu^{n+1}| |\partial^\alpha D\mu^n| \, dx d\tau.$$

- We use the assumption **(H1)**:

$$I_A \leq \varepsilon F(M_n) \int_0^T \int_{\mathbb{T}^d} |\partial^\alpha \mu^{n+1}| |\partial^\alpha D\mu^n| \, dx d\tau.$$

- Next we use Young's inequality, with parameter $\sigma_1 > 0$.

The term I_A (continued)

- Using Young's inequality yields the following:

$$\begin{aligned} I_A &\leq \varepsilon F(M_n) \left(\frac{1}{2\sigma_1} \int_0^T \|\partial^\alpha \mu^{n+1}\|_0^2 d\tau + \frac{\sigma_1}{2} \int_0^T \|\partial^\alpha D\mu^n\|_0^2 d\tau \right) \\ &\leq \varepsilon F(M_n) \left(\frac{1}{2\sigma_1} \int_0^T \|\partial^\alpha \mu^{n+1}\|_0^2 d\tau + \frac{\sigma_1}{2} N_n \right). \end{aligned}$$

- We choose $\sigma_1 = 28T\varepsilon F(M_n)$. This implies

$$\begin{aligned} I_A &\leq \frac{1}{56T} \int_0^T \|\partial^\alpha \mu^{n+1}\|_0^2 d\tau + 14\varepsilon^2 T (F(M_n))^2 N_n \\ &\leq \frac{1}{56} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 \right) + 14\varepsilon^2 T (F(M_n))^2 N_n. \end{aligned}$$

The term I_B

- We have the definition

$$I_B = \varepsilon \int_0^t \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}).$$

$$[(\partial^\alpha D\mu^n) \cdot \Theta_p(\tau, x, \mu^n, Dw^n) - \partial^\alpha (D\mu^n \cdot \Theta_p(\tau, x, \mu^n, Dw^n))] dx d\tau.$$

- We use Young's inequality with parameter $\sigma_2 = 28T\varepsilon$, finding

$$I_B \leq \frac{1}{56} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 \right) + 14\varepsilon^2 T \int_0^T \left\| (\partial^\alpha D\mu^n) \cdot \Theta_p(\tau, \cdot, \mu^n, Dw^n) - \partial^\alpha (D\mu^n \cdot \Theta_p(\tau, \cdot, \mu^n, Dw^n)) \right\|_0^2 d\tau.$$

- We then use our lemma about derivatives of products, and Sobolev embedding, and **(H1)**, finding

$$I_B \leq \frac{1}{56} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 \right) + c\varepsilon^2 T^2 (F(M_n))^2 M_n (1 + M_n)^{s-1}.$$

The term II

- We have the definition

$$II = -\varepsilon \int_0^t \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) \partial^\alpha \left((\mu^n + \bar{m}) \sum_{i=1}^d \Theta_{x_i p_i}(\tau, x, \mu^n, Dw^n) \right) dx d\tau$$

- There is no need to extract a leading-order contribution this time.
- We use Young's inequality, the Sobolev algebra property, and **(H1)**, finding

$$II \leq \frac{1}{56} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 \right) + c\varepsilon^2 T^2 (F(M_n))^2 (1 + M_n)^s.$$

The term III_A

- We have the definition

$$III_A = -\varepsilon \int_0^t \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) (\mu^n + \bar{m}) \cdot \sum_{i=1}^d [(\Theta_{qp_i}(\tau, x, \mu^n, Dw^n)) (\partial^\alpha \partial_{x_i} \mu^n)] dx d\tau.$$

- We estimate this by putting μ^n and Θ_{qp_i} in L^∞ using Sobolev embedding and **(H1)**, and we use N_n for $\partial^\alpha \partial_{x_i} \mu^n$.
- The result is

$$III_A \leq \frac{1}{56} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 \right) + c\varepsilon^2 T (1 + M_n) (F(M_n))^2 N_n.$$

The term III_B

- We have the definition

$$III_B = -\varepsilon \int_0^t \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) \left\{ \partial^\alpha \left((\mu^n + \bar{m}) \sum_{i=1}^d [(\Theta_{qp_i}(\tau, x, \mu^n, Dw^n)) (\partial_{x_i} \mu^n)] \right) \right. \\ \left. - (\mu^n + \bar{m}) \sum_{i=1}^d [(\Theta_{qp_i}(\tau, x, \mu^n, Dw^n)) (\partial^\alpha \partial_{x_i} \mu^n)] \right\} dx d\tau.$$

- We estimate this similarly to how we estimated I_B :

$$III_B \leq \frac{1}{56} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 \right) + cT^2 \varepsilon^2 (F(M_n))^2 (1 + M_n)^{s+1}.$$

The term IV_A

- We have the definition

$$IV_A = -\varepsilon \int_0^t \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) (\mu^n + \bar{m}) \cdot \sum_{i=1}^d \sum_{j=1}^d \left[(\Theta_{p_i p_j}(\tau, x, \mu^n, Dw^n)) (\partial^\alpha \partial_{x_i x_j}^2 w^n) \right] dx d\tau.$$

- As in III_A , we put μ^n and $\Theta_{p_i p_j}$ in L^∞ , using Sobolev embedding and **(H1)**, and we use N_n to bound $\partial^\alpha \partial_{x_i x_j}^2 w^n$:

$$IV_A \leq \frac{1}{56} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 \right) + c\varepsilon^2 T (1 + M_n) (F(M_n))^2 N_n.$$

The term IV_B

- We have the definition

$$IV_B = -\varepsilon \int_0^t \int_{\mathbb{T}^d} (\partial^\alpha \mu^{n+1}) \left\{ \partial^\alpha \left((\mu^n + \bar{m}) \sum_{i=1}^d \sum_{j=1}^d \left[(\Theta_{p_i p_j}(\tau, x, \mu^n, Dw^n)) (\partial_{x_i x_j}^2 w^n) \right] \right) \right. \\ \left. - (\mu^n + \bar{m}) \sum_{i=1}^d \sum_{j=1}^d \left[(\Theta_{p_i p_j}(\tau, x, \mu^n, Dw^n)) (\partial^\alpha \partial_{x_i x_j}^2 w^n) \right] \right\} dx d\tau.$$

- We estimate this similarly to I_B and III_B :

$$IV_B \leq \frac{1}{56} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 \right) + cT^2 \varepsilon^2 (F(M_n))^2 (1 + M_n)^{s+1}.$$

What have we been doing?

- We need to bound the growth of the approximate solutions.
- We have shown how to bound $\mu^{n+1,\delta}$, mainly in terms of the previous iterates.
- The structure in the equations that we have used: considering the MFG system as a system for (m, Du) , other than the linear terms, the equations only have first spatial derivatives. The linear terms are parabolic, giving a gain of regularity of one derivative. We have used this gain of regularity to control the first spatial derivatives present in the nonlinear terms.

Using our estimate for $\mu^{n+1,\delta}$

- We have shown

$$I + II + III + IV \leq \frac{1}{8} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 \right) + c\varepsilon^2 T (F(M_n))^2 \left((1 + T)(1 + N_n)(1 + M_n)^{s+1} \right).$$

- This has as one consequence

$$\frac{1}{2} \|\partial^\alpha \mu^{n+1}(t, \cdot)\|_0^2 \leq \frac{1}{2} \|\partial^\alpha \mu^{n+1}(0, \cdot)\|_0^2 + \frac{1}{8} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 \right) + c\varepsilon^2 T (F(M_n))^2 \left((1 + T)(1 + N_n)(1 + M_n)^{s+1} \right).$$

Using our estimate for $\mu^{n+1,\delta}$ (continued)

- Another consequence is

$$\int_0^t \|D\partial^\alpha \mu^{n+1}\|_0^2 d\tau \leq \frac{1}{2} \|\partial^\alpha \mu^{n+1}(0, \cdot)\|_0^2 + \frac{1}{8} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 \right) + c\varepsilon^2 T (F(M_n))^2 \left((1+T)(1+N_n)(1+M_n)^{s+1} \right).$$

- We take the supremum in time of both of these consequences, and we add, finding

$$\frac{1}{4} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 \right) + \int_0^T \|D\partial^\alpha \mu^{n+1}\|_0^2 d\tau \leq \|\partial^\alpha \mu^{n+1}(0, \cdot)\|_0^2 + c\varepsilon^2 T (F(M_n))^2 \left((1+T)(1+N_n)(1+M_n)^{s+1} \right).$$

We do the same thing for $w^{n+1,\delta}$

- We go through the same steps for the w equation, and we estimate $V, VI_A, VI_B, VII_A,$ and VII_B .
- We get a corresponding conclusion for w and we add it to the conclusion for μ , finding

$$\begin{aligned} & \frac{1}{4} \left(\sup_{t \in [0, T]} \|\partial^\alpha \mu^{n+1}\|_0^2 + \sup_{t \in [0, T]} \|\partial^\alpha \partial_{x_j} w^{n+1}\|_0^2 \right) \\ & + \int_0^T \|D\partial^\alpha \mu^{n+1}(\tau, \cdot)\|_0^2 d\tau + \int_0^T \|D\partial^\alpha \partial_{x_j} w^{n+1}(\tau, \cdot)\|_0^2 d\tau \\ & \leq \|\partial^\alpha \mu^{n+1}(0, \cdot)\|_0^2 + \|\partial^\alpha \partial_{x_j} w^{n+1}(T, \cdot)\|_0^2 \\ & \quad + c\varepsilon^2 T (F(M_n))^2 \left((1+T)(1+N_n)(1+M_n)^{s+1} \right). \end{aligned}$$

Ready for our induction

- We sum over multi-indices and whatnot, we multiply by 4, and we substitute the boundary conditions:

$$M_{n+1} + 4N_{n+1} \leq 4\|\mathbb{P}_\delta \mu_0\|_{s-1}^2 + 4\|D\mathbb{P}_\delta w_T\|_{s-1}^2 + c\varepsilon^2 T(F(M_n))^2 \left((1+T)(1+N_n)(1+M_n)^{s+1} \right). \quad (4)$$

- Let \mathcal{S} be a real number such that

$$4\|\mu_0\|_{s-1}^2 + 4\|Dw_T\|_{s-1}^2 \leq \mathcal{S}.$$

- We state our smallness assumption:

(H2) The function F and the constants c , ε , T , and \mathcal{S} satisfy

$$c\varepsilon^2 T(F(2\mathcal{S}))^2 \left((1+T)(1+2\mathcal{S})^{s+2} \right) \leq \mathcal{S}.$$

Induction proof: the end of the estimate

- Claim: for all n , we have $M_n + 4N_n \leq 2\mathcal{S}$.
- Base case: $\mu^0 = w^0 = 0$, so $M_0 + N_0 = 0$.
- Inductive hypothesis: Assume $M_n + 4N_n \leq 2\mathcal{S}$.
- Notice that $M_n \leq 2\mathcal{S}$ and $N_n \leq 2\mathcal{S}$.
- Also, recall that $\|\mathbb{P}_\delta f\|_{s-1} \leq \|f\|_{s-1}$, for any f .
- We then have

$$M_{n+1} + 4N_{n+1} \leq 4\|\mu_0\|_{s-1}^2 + 4\|Dw_T\|_{s-1}^2 + c\varepsilon^2 T(F(2\mathcal{S}))^2 \left((1+T)(1+2\mathcal{S})^{s+2} \right). \quad (5)$$

- So, $M_{n+1} + 4N_{n+1} \leq \mathcal{S} + \mathcal{S} = 2\mathcal{S}$.
- This completes the proof.

About the smallness constraint

- Let's state that smallness constraint again: \mathcal{S} is bigger than a multiple of the data, and we assume

$$c\varepsilon^2 T (F(2\mathcal{S}))^2 \left((1+T)(1+2\mathcal{S})^{s+2} \right) \leq \mathcal{S}.$$

- Given a Hamiltonian and some data, and a time horizon T , this can be enforced by taking ε small.
- Given a Hamiltonian and some data and a value of ε , then this can be enforced by taking T small.
- For some Hamiltonians, it could be the case that $(F(2\mathcal{S}))^2 \rightarrow 0$ as $\mathcal{S} \rightarrow 0$, faster than \mathcal{S} . Then, given values of ε and T , the smallness constraint can be enforced by taking the data sufficiently small.
- So, the smallness constraint simultaneously considers the length of the time horizon, the strength of the coupling in the system, and the size of the data.

Thanks for your attention.