

LECTURE 22, May 30, 2023

MEAN-FIELD GAMES.

PROBLEM: Hard to study NASH EQUILIBRIA with MANY players.

If we look for equil. in feedback form via PDES: state \rightarrow d state dim. for each agent, N agents. Full state: dim Nd
system of N PDES, each in \mathbb{R}^{Nd}

Examples . financial markets $N \sim 10^4$

• ENERGY DISTRIBUTION. $N \sim 10^5$

• CROWD MOTION $N \sim 10^2$

Solving N PDES in Nd dim is NOT FEASIBLE if $Nd > 10$.

Idea. 2006 J. Lasry, P.L. Lions

2005-6 P. Caines, M. Huang, R. Malhan

Supp.: "agents are similar", and they interact only via the costs & via the EMPIRICAL MEASURE of the other players $(i \neq j)$: $\frac{1}{N-1} \sum_{i \neq j} \delta_{x_i}$

Similarity with MEAN-FIELD THEORIES in PHYSICS:

Look for simpler MACROSCOPIC DESCRIPTION of complex phenomena instead MICROSCOPIC description.

Q.: How describe populations of particles or rational agents?

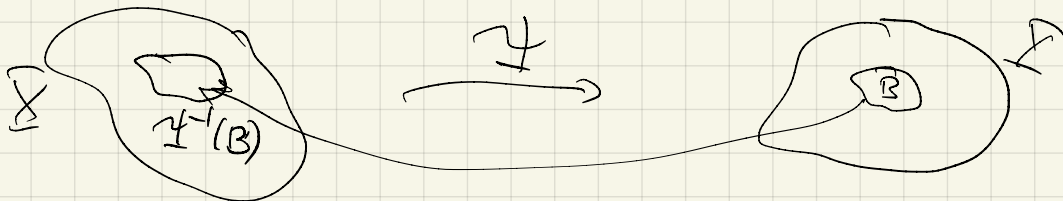
Answer.: by its distribution in the state space:

$$\{ \text{amount of agents in } B \subseteq \mathbb{R}^d \text{ at time } t \} = \mu_t(B) = \int_B d\mu_t(x) \stackrel{\uparrow}{=} \int_B m(x,t) dx$$

if μ_t has a density $m(\cdot, t)$

Q.: How does μ_t evolve in time?

Def.: $\Psi: \underline{X} \rightarrow \underline{Y}$, $\underline{X}, \underline{Y}$ metric, $\mu \in \mathcal{P}(\underline{X})$



The PUSH-FORWARD measure of μ via Ψ is: $\Psi\# \mu \in \mathcal{P}(\underline{Y})$

Def

$$(\#) \quad \Psi\# \mu(B) := \mu(\Psi^{-1}(B))$$

$$\chi_Z(y) := \begin{cases} 1 & y \in Z \\ 0 & y \notin Z \end{cases}$$

$$\underline{R}unh \quad (\#) \Leftrightarrow \int_{\underline{Y}} \chi_B(y) d(\Psi\# \mu)(y) := \int_{\underline{X}} \chi_{\Psi^{-1}(B)}(x) d\mu(x)$$

$$= \int_{\underline{X}} \chi_B(\Psi(x)) d\mu(x)$$

Any $f: \underline{Y} \rightarrow \mathbb{R}$ meas & can be approximated by

"simple functions" \Rightarrow

$$\int_{\Sigma} g(y) d(\Psi \# \mu)(y) = \int_{\Sigma} g(\Psi(x)) d\mu(x). \quad (*)$$

Q: How does μ_t evolve if each agent follows $\dot{y}(s) = f(y(s), s)$ (DS)?

Def. Φ = flow ass. to (DS) i.e.

$$\Phi(x, s) = \text{sol. at time } s \text{ of } \begin{cases} \dot{y} = f(y, s) \\ y(0) = x \end{cases}$$

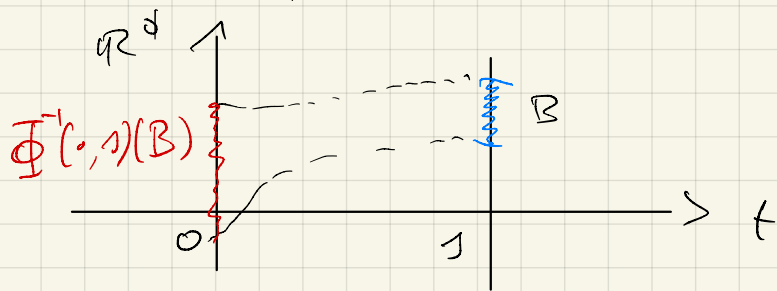
$$\Rightarrow \frac{\partial \Phi}{\partial s} = f(\Phi(x, s), s), \quad \Phi(x, 0) = x \quad \forall x$$

N.B. f Lipitz y , meas. in s \Rightarrow $x \mapsto \Phi(x, s)$ is BIJECTIVE

its inverse is $\Phi^{-1}(\cdot, s)$

Def. Given $\mu_0 \in \mathcal{P}(\mathbb{R}^d)$, $\mu_s := \Phi(\cdot, s) \# \mu_0$ is

push forward of μ_0 by Φ (or by f, \dots)



Want to derive an equation for μ_s .

Take $\psi: \mathbb{R}^d \times [0, T] \rightarrow \mathbb{R}$ meas. & (test) fn. Ass. $\psi \in C^1$

$$(*) \Rightarrow \int_{\mathbb{R}^d} \psi(y, s) d\mu_s(y) \stackrel{(+)}{=} \int_{\mathbb{R}^d} \psi(\Phi(x, s), s) d\mu_0(x)$$

$\Phi \# \mu_0$ $\mu_s = \Phi(\cdot, s) \# \mu_0$

$$\frac{d}{ds} \int_{\mathbb{R}^d} \psi(y, s) d\mu_s(y) = \int_{\mathbb{R}^d} \frac{d}{ds} \psi(\Phi(x, s), s) d\mu_0(x) =$$

$$= \int_{\mathbb{R}^d} \left(\frac{\partial \psi}{\partial s} + \nabla_x \psi \cdot \Phi_s \right) (\Phi(x, s), s) d\mu_0(x) \stackrel{(+)}{=}$$

"f"

$$= \int_{\mathbb{R}^d} (\psi_s + \nabla_x \psi \cdot f)(y, s) d\mu_s(y)$$

Now sum. Supp ψ compact in $\mathbb{R}^d \times [0, T]$

$$\int_0^T \boxed{\phantom{\int_{\mathbb{R}^d} \psi(y, s) d\mu_s(y)}} \Rightarrow$$

$$\int_{\mathbb{R}^d} \psi(y, s) d\mu_s(y) \Big|_{s=0}^{s=T} = - \int_{\mathbb{R}^d} \psi(y, s) d\mu_0(y) =$$

$$= \int_0^T \int_{\mathbb{R}^d} (\psi_s + \nabla_x \psi \cdot f)(y, s) d\mu_s(y) \quad \forall \psi \in C_c^1(\mathbb{R}^d \times [0, T])$$

(WCE)

Def. (WCE) is weak form (distributional) of the CONTINUITY EQUATION

$$(CE) \quad \frac{\partial m}{\partial s} + \operatorname{div}_x (m f) = 0$$

with initial condition μ_0 .

We proved:

Lemma. Push-forward $\mu_s = \Phi(\cdot, s) \# \mu_0$
of $\mu_0 \in \mathcal{P}(\mathbb{R}^d)$ via f satisfies (WCE).

Motivation of the name "continuity eq.":

S-uv. μ_s sol. of (WCE) has density:

$$d\mu_0(x) = m_0(x) dx, \quad d\mu_s(x) = m(x, s) dx \quad \&$$

$$m \in C'(\mathbb{R}^d \times (0, T)) \cap C(\mathbb{R}^d \times [0, T]).$$

CLAIM: m solves (CE). (WCE) is

$$0 = \int_{\mathbb{R}^d} \psi(y, 0) m_0(y) dy + \int_0^T \int_{\mathbb{R}^d} (\psi_s + D_x \psi \cdot f) m(y, s) dy ds \quad = (I)$$

$$\int_0^T (\psi_s m)(y, s) ds = \psi m \Big|_0^T - \int_0^T \psi m_s ds =$$

Integrate by parts \rightarrow $= - \int_{\mathbb{R}^d} \psi(y, 0) m_0(y) dy - \int_0^T (\psi m_s)(y, s) ds$

$$\operatorname{div}_x (m \psi f) = \operatorname{div}_x (m f) \psi + D_x \psi \cdot (m f)$$

$$(I) = \int_0^T \int_{\mathbb{R}^d} [(-m_s - \operatorname{div}_x (m f)) \psi + \operatorname{div}_x (m \psi f)] dy ds$$

Gauss Th.

$$\int_{\mathbb{R}^d} \operatorname{div}_x (m \psi f) dy = \int_{\partial \operatorname{supp} \psi} m \psi f \cdot \nu d\sigma = 0$$

\uparrow ext. normal.

$$\Rightarrow 0 = \int_0^T \int_{\mathbb{R}^d} (m_s + \operatorname{div}_x (m f)) \psi dy ds \quad \forall \psi$$

By the arbitrariness of $\varphi \Rightarrow u_\varphi + \operatorname{div}_x (u \varphi) = 0$
in \mathbb{R}^d .

Viceversa (HW) If u solves (CF) & attains initial data, then $d\varphi_s = u(-, s) dx$ satisfies (WCE)

A HEURISTIC DERIVATION of the MFC (mean field game) System of PDES

Ref. • Cardaliaguet LN 2013 (also with Porretta...)
• P. L. Lions lectures at Collège de France.

Take a population of agents with dynamics
 $\dot{y}(s) = -a(s)$, $a(s) \in \mathbb{R}^d$, $y(t) = x$

Cost functional of the generic agent :

$$J(t, x, a(\cdot)) := \int_0^t (L(a(s)) + F(y(s), s)) ds + g(y(T))$$

Ass. $\lim_{|a| \rightarrow +\infty} \frac{L(a)}{|a|} = +\infty$, L convex, $H = L^*$ convex
 $\text{co} L'$.

HJB associated is

$$(HJB) \begin{cases} -u_t + H(D_x u) = F(x, t) & \text{in } \mathbb{R}^d \times (0, T) \\ u(x, T) = g(x) \end{cases}$$

Lemma (Verif. thm. \approx Cor. in Lect. 5)

If $u \in C^1$ solves (HJB), LeC^1 , DL invertible, $(DL)^{-1}eC^1$
 $\Rightarrow DH(D_x u(x, t))$ is an optimal feedback, i.e.,

$$\left\{ \begin{array}{l} \dot{y}(s) = - \underbrace{DH(D_x u(y(s), s))}_{=: \hat{a}^*(s)} \quad s > t \\ y(t) = x \end{array} \right.$$

Let a sol. $\hat{a}^*(\cdot)$ minimize $J(t, x, \cdot)$.

Consequence of (CE) & (HJB): If all agents have the same cost $L + F$, are all "rational", & $J \in C^1$ sol of (HJB), then the distribution of the population is a weak solut. of.

$$\left\{ \begin{array}{l} m_t - \operatorname{div}_x(m DH(D_x u)) = 0 \\ m(x, 0) = m_0(x) \end{array} \right.$$

Now supp. for N players, the cost of the N -th depends also on the empirical measure

$$\mu^N(s) = \frac{1}{N-1} \sum_{i=1}^{N-1} \delta_{y^i(s)}$$

$\delta = \text{Dirac}$
 $y^i(s) = \text{position of } i\text{-th player}$

$$F: \mathbb{R}^d \times \mathcal{P}(\mathbb{R}^d) \rightarrow \mathbb{R}$$

$$(y^N(s), \mu^N(s))$$

The value function v^N of the N -th player solves

$$\left\{ \begin{array}{l} - \frac{\partial v^N}{\partial t} + H(D_x v^N) = F(x, \mu^N(t)) \\ v^N(x, T) = g(x) \end{array} \right.$$



Supp. all players behave OPTIMALLY, all equal to the N th. Then $\rho_{y(i)}$ is a weak sol. of

$$\begin{cases} \rho_t - \operatorname{div}(\rho DH(Dv^N)) = 0 \\ \rho_0 = \delta_{x^i} \end{cases}, \text{ then}$$

$$\begin{cases} \frac{\partial \rho^N}{\partial t} - \operatorname{div}(\rho^N DH(Dv^N)) = 0 \\ \rho^N(x, 0) = \rho_0(x) \end{cases} \quad \leftarrow$$

Assume as $N \rightarrow \infty$ $\rho_t^N \rightarrow \rho_t \forall t$, ρ_t has a density $m(\cdot, t)$, $v^N \rightarrow u$ with all derivatives, F cont.

I "expect":

$$(MFG) \begin{cases} -u_t + H(D_x u) = F(x, m) & \text{in } \mathbb{R}^d \times (0, T) \\ m_t - \operatorname{div}(m DH(D_x u)) = 0 & \text{"} \\ m(x, 0) = m_0(x), \quad u(x, T) = g(x). \end{cases}$$

N.B. system of backward HJB eq.

• forward cont. eq.

A rigorous proof of the limit $N \rightarrow \infty$ is not known yet.

Can give a GAME-THEORETIC justification of (MFG) as describing a Nash-type equilibrium,

Def. A pair (μ, u) , $\mu: [0, T] \rightarrow \mathcal{P}(\mathbb{R}^d)$,
 $u: \mathbb{R}^d \times [0, T] \rightarrow \mathbb{R}$ is a MFGs equil. if.

- u is the value fun. of the opt. contr. probl. with cost $L(x(s)) + F(y(s), \mu_s)$
- μ is the distr's. of a population of players all following the feedback $DH(D_x u)$, optimal for previous contr. pl.
- "For an agent it is not convenient to deviate from $DH(D_x u)$ if the rest of the population does not deviate".