

Set Values of Mean Field Games

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- 1 Set Valued Frameworks
- 2 Set Values of Mean Field Games
 - Discrete Setup
 - Stability & Sensitivity of Games
 - Results
- 3 Diffusion Model

1 Set Valued Frameworks

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3 Diffusion Model

- \mathbb{V} : parameters (t, x, μ, \dots) \rightarrow collections
 - Geometric Surface Evolutions $\left[\partial_t \mathbb{V}(t) = h(t, y, \mathbf{n}, \partial_y \mathbf{n}) \right]$
 - Mean curvature flows, crystal formations, image processing
 - Stochastic Viability & Target Problems
 - Dynamic Risk Measures
 - N -player Games
 - Multivariate Control Problems
 - Mean-field Games

Multivariate Control Problem

$$\mathbb{V}(t, x) := \left\{ J(t, x, \alpha) : \forall \alpha \right\} \in \mathbb{R}^m$$

- Dynamic Programming Principle
i.e. Time-consistency
- Itô formula
- PDE (Hamilton-Jacobi-Bellman)

Mean-field Games

$$\mathbb{V}(t, \mu) := \left\{ J(t, \mu, \alpha) : \forall \alpha \text{ equilibrium} \right\}$$

- Dynamic Programming Principle
- Convergence of \mathbb{V}^N to \mathbb{V}
- PDE (?) (Master Equation)

N-Player Game

– Actions

α^i : controls of the player

– Dynamics

X_t^i : state of individual player

$$\mu_t^N = \frac{1}{N} \sum_j \delta_{X_t^j}$$

– Cost

$J_i(\vec{\alpha})$: cost of the player

– Equilibrium

Nash Equilibrium:

$$J_i(\vec{\alpha}^*) \leq J_i(\vec{\alpha}^*, \alpha^i)$$

Mean Field Game

– Actions

α : control of the population

$\tilde{\alpha}$: control of a player

– Dynamics

X_t^α : state of the population

$$\mu_t^\alpha = \mathcal{L}_{X_t^\alpha}$$

$X_t^{\alpha, \tilde{\alpha}}$: state of a player

– Cost

$J(\alpha, \tilde{\alpha})$: cost of a player

– Equilibrium

Mean field Equilibrium:

$$J(\alpha^*, \alpha^*) \leq J(\alpha^*, \tilde{\alpha})$$

Examples to Equilibria without Dynamics

N-Player Game

$A = \{0, 1\}$, two player.

(J_1, J_2)	$a_1 = 0$	$a_1 = 1$
$a_2 = 0$	(2,2)	(3,1)
$a_2 = 1$	(1,3)	(1,1)

Mean Field Game

$A = [0, 1]$,

$$J(a, \tilde{a}) = 1 - a\tilde{a}$$

$a = 0$ and $a = 1$ are equilibria.

Examples of Different Structures

- Games on sparse & dense graphs
 - Games of timing (optimal stopping)
 - Cooperative Equilibriums
 - Games with a major player
 - Games with clusters
 - Ergodic games
 - Finite player games
 - \vdots
- Finite/continuous state/time
 - Common noise
 - Diffusion with jumps
 - Information available to players
 - State/path dependent dynamics
 - State/path dependent costs
 - Non-symmetric costs for players
 - \vdots

Classical References

Huang-Malhamé-Caines (2006)

Lasry-Lions (2007)

Lions (2008)

Cardaliaguet (2010)

Carmona-Delarue (2018)

⋮

References on Master Equation Approach (under monotonicity condition)

Buckdahn-Li-Peng-Rainer (2017)

Chassagneux-Crisan-Delarue (2014)

Cardaliaguet-Delarue-Lasry-Lions (2019)

Gangbo-Mészáros-Mou-Zhang (2021)

⋮

Objective & Related Literature

Our objective is to study dynamic set value $\mathbb{V}(t, \mu)$:

- (i) Establish time-consistency (DPP)
- (ii) characterize \mathbb{V} as an appropriate limit of \mathbb{V}^N (Convergence)

- Set Value is by definition unique and exists.
- We consider convergence of Set Values instead of individual equilibriums.

Settings

- Discrete time & space | state dependent controls | homogeneous controls
- Discrete time & space | path dependent controls | heterogeneous controls
- Continuous time & space | state dependent controls | homogeneous controls

Related to DPP

Feinstein-Rudloff-Zhang (2020)

Related to Convergence

Cardaliaguet-Delarue-Lasry-Lions (2019)

Lacker (2014, 2020)

Lacker-Flem (2021)

Possamai-Tangpi (2021)

Djete (2021)

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Discrete State Dependent Setup

- \mathbb{S} , finite state space.
- $\mathbb{T} = \{0, \dots, T\}$, finite time steps.
- \mathbb{A} is the set of possible values of the control.
- Transition function:

$$q(t, x, \nu, a; \tilde{x}) : \mathbb{T} \times \mathbb{S} \times \mathcal{P}(\mathbb{S}) \times \mathbb{A} \times \mathbb{S} \rightarrow (0, 1]$$

$$\text{where } \sum_{\tilde{x} \in \mathbb{S}} q(t, x, \nu, a; \tilde{x}) = 1$$

Given $(t, \mu, \alpha; \tilde{\alpha}, x)$,

$$\text{(Population)} \quad \mu_{s+1}^\alpha(\tilde{x}) = \sum_{x \in \mathbb{S}} \mu_s^\alpha(x) q(s, x, \mu_s^\alpha, \alpha(s, x, \mu_s^\alpha); \tilde{x})$$

$$\text{(Player)} \quad \mathbb{P}^{\mu^\alpha; t, x, \tilde{\alpha}}(X_{s+1} = x | X_s) = q(s, X_s, \mu_s^\alpha, \tilde{\alpha}(s, X_s, \mu_s^\alpha); x)$$

$$J(t, \mu, \alpha; \tilde{\alpha}, x) \doteq \mathbb{E}^{\mathbb{P}^{\mu^\alpha; t, x, \tilde{\alpha}}} \left[G(X_T, \mu_T^\alpha) + \sum_t^{T-1} F(\dots) \right]$$

Definition

We say $\alpha^* \in \mathcal{M}(t, \mu)$ a MFG Equilibrium if

$$J(t, \mu, \alpha^*; \alpha^*, x) \leq J(t, \mu, \alpha^*; \tilde{\alpha}, x), \quad \forall x, \tilde{\alpha}$$

Definition

$$\mathbb{V}(t, \mu) = \left\{ J(t, \mu, \alpha^*; \alpha^*, \cdot) : \text{for all } \alpha^* \in \mathcal{M}(t, \mu) \right\}$$

Given $(t, \vec{x}, \vec{\alpha})$,

$$\mathbb{P}^{t, \vec{x}, \vec{\alpha}}(\vec{X}_{s+1} = \vec{x}' | \vec{X}_s) = \prod_j q(s, X_s^j, \mu_{\vec{X}_s}^N, \alpha^j(X_s^j, \mu_{\vec{X}_s}^N); x_j')$$

$\vec{X} = (X^1, \dots, X^N)$ is the canonical process on $(\mathbb{T} \times \mathbb{S})^N$.

$$J_i^N(t, \vec{x}, \vec{\alpha}) \doteq \mathbb{E}^{\mathbb{P}^{t, \vec{x}, \vec{\alpha}}} \left[G(X_T^i, \mu_{\vec{X}_T}^N) + \sum_t^{T-1} F(\dots) \right]$$

Definition

We say $\vec{\alpha}^* \in \mathcal{M}^N(t, \vec{x})$ if

$$J_i^N(t, \vec{x}, \vec{\alpha}^*) \leq J_i^N(t, \vec{x}, \vec{\alpha}^{*, -i} \tilde{\alpha}^i), \quad \forall i, \tilde{\alpha}^i$$

Definition

$$\mathbb{V}^N(t, \mu_{\vec{x}}^N) \doteq \left\{ (J_i^N(t, \vec{x}, \vec{\alpha}^*))_i : \text{for all } \vec{\alpha}^* \in \mathcal{M}^N(t, \vec{x}) \right\}$$

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ϵ -MFG Equilibrium: $J(t, \mu, \alpha^*; \alpha^*, x) - \epsilon \leq J(t, \mu, \alpha^*; \tilde{\alpha}, x)$.

Define the Set Value as;

$$\mathbb{V}(t, \mu) \doteq \bigcap_{\epsilon > 0} \mathbb{V}_\epsilon(t, \mu)$$

- This definition is analogous to the standard control theory, as the value defined to be the infimum over ϵ -optimal controls:

$$v(t, x) = \lim_{\epsilon \rightarrow 0} J_\epsilon(t, x), \quad J_\epsilon(t, x) = \inf_{\alpha^\epsilon} J(t, x, \alpha^\epsilon)$$

- It is possible that there is no optimal control.

N-Player Game

$\mathbb{A} = \{0, 1\}$, two player.

(J_1, J_2)	$a_1 = 0$	$a_1 = 1$
$a_2 = 0$	(2,2)	(3,1- δ)
$a_2 = 1$	(1- δ ,3)	(1,1)

Mean Field Game

$\mathbb{A} = [0, 1]$,

$$J(a, \tilde{a}) = 1 - (a + \delta)\tilde{a}$$

Only $a = 1$ is an equilibrium.

- Extension to ϵ -equilibria is crucial to have stability hence convergence.

Information that control depends on

- $\alpha = \alpha(t, X_t^i, \mu_t^N)$

State Dependent

[DPP holds / Convergence holds]

- $\alpha = \alpha(t, X_{[0,t]}^i, \mu_{[0,t]}^N)$

Path Dependent

[DPP holds / Convergence holds (Continuous?)]

- $\alpha = \alpha(t, X_1, \dots, X_N)$

Full information (\rightarrow weak MFG)

- ...

For games, difference is crucial because when equilibrium is not unique, set values depend on the choice.

- $\mathbb{V}_{state} \neq \mathbb{V}_{path}$ •

Notes on Convergence

- In MFG, there is only one control that population use.

However, we need 'same' type of controls for both NPG and MFG:

- $\mathbb{V}(t, \mu)$ is characterized as the limit fo homogeneous controls
($\alpha^1 = \dots = \alpha^N$)
- Without this restriction, limit of NPG is characterized by using relax control for MFG. ($\int_{\mathbb{A}} q(s, x, \mu^\gamma, a; \bar{x}) \gamma(s, x, da)$)

- Introduce corresponding Set Values;

$$\mathbb{V}_\epsilon^{N, hom}(t, \mu_{\bar{x}}^N) \quad \text{and} \quad \mathbb{V}_\epsilon^{relax}(t, \mu)$$

An Important Observation

Introduce

$$\Lambda^N(t, \vec{x}, \vec{\alpha}) = \frac{1}{N} \sum_{j=1}^N \delta_{(x^j, \alpha^j)} \in \mathcal{P}(\mathbb{X}_t \times \mathcal{A}_{path})$$

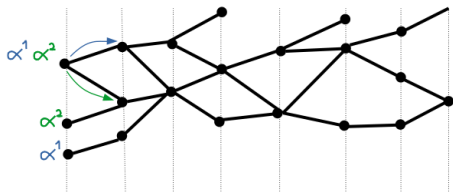
There exists J^N such that

$$J_i^N(t, \vec{x}, \vec{\alpha}) = J^N(t, \Lambda^N(t, \vec{x}, \vec{\alpha}); x^i, \alpha^i)$$

Next, we introduce MFG corresponding to $\Lambda \in \mathcal{P}(\mathbb{X}_t \times \mathcal{A}_{path})$ with marginal and conditional distribution;

$$\mu^\Lambda(x) \doteq \int_{\mathcal{A}_{path}} \Lambda(x, d\alpha), \quad \Lambda_{x.}(\alpha) \doteq \frac{\Lambda(x, \alpha)}{\mu^\Lambda(x)}$$

Global Formulation for MFG



MFG: Given $(t, \Lambda; \tilde{\alpha}, x)$, introduce a mapping $\Lambda_s : \mathbb{T}_t \rightarrow \mathcal{P}(\mathbb{X} \times \mathcal{A}_{path})$,

$$\text{(Population)} \quad \Lambda_{s+1}(x, \alpha) \doteq q(s, x_s, \mu_s^\Lambda, \alpha(s, x, \mu^\Lambda); x_{s+1}) \Lambda_s(x, \alpha)$$

$$\text{(Player)} \quad \mathbb{P}^{\mu^\Lambda; t, x, \tilde{\alpha}}(X_{s+1} = x | X_\cdot) = q(s, X_s, \mu_s^\Lambda, \tilde{\alpha}(s, X_\cdot, \mu_s^\Lambda); x)$$

$$J(t, \Lambda; \tilde{\alpha}, x) = \mathbb{E}^{\mathbb{P}^{\mu^\Lambda; t, x, \tilde{\alpha}}} \left[G(X_T, \mu_T^\Lambda) + \sum_{s=t}^{T-1} F(\dots) \right]$$

Global Formulation and Relax Controls

MFg

population $\alpha \in A$
player $\tilde{\alpha} \in A$

 $\mathcal{J}(\alpha^*; \alpha^*) \geq \mathcal{J}(\alpha^*, \alpha)$

 α^* equilibrium

Relaxed

— $\gamma \in \mathcal{P}(A)$
— $\tilde{\gamma} \in \mathcal{P}(A)$

 $\mathcal{J}(\gamma^*; \gamma^*) \geq \mathcal{J}(\gamma^*, \gamma)$

 γ^* equilibrium

Global

— $\Gamma \in \mathcal{P}(A)$
— $\alpha \in A$

 $\mathcal{J}(\Gamma; \alpha) \geq \mathcal{J}(\Gamma; \tilde{\alpha})$

 Γ equilibrium
 $\alpha \in \text{supp } \Gamma$

- NPG is naturally connected with $\Lambda^N = (1/N) \sum_j \delta_{(x^j, \alpha^j)}$.

Theorem

$$\mathbb{V}^{\text{global}}(t, \mu) = \mathbb{V}^{\text{relax}}(t, \mu)$$

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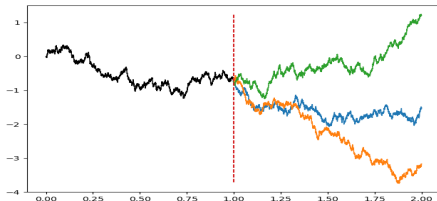
Time Consistency - Dynamic Programming Principle

- Define cost function up to time $T_0 \leq T$ with a given terminal cost ψ .
- Introduce appropriate MFG Equilibrium $\mathcal{M}(T_0, \psi, t, \mu)$.

Theorem

$$\mathbb{V}(t, \mu) = \left\{ J(T_0, \psi; t, \mu, \alpha^*; \alpha^*, \cdot) : \right.$$

$\left. \text{for some } \alpha^* \in \mathcal{M}(T_0, \psi, t, \mu) \text{ and } \psi \in \mathbb{V}(T_0, \mu_{T_0}^{\alpha^*}) \right\}$



- Holds for \mathbb{V}^{raw} , \mathbb{V} , \mathbb{V}^{relax} .
- DPP is central for PDE approach, which is an ongoing project.

Theorem

For $\mu_{\vec{x}}^N \Rightarrow \mu$,

$$\bigcap_{\epsilon > 0} \liminf_{N \rightarrow \infty} \mathbb{V}_{\epsilon}^{N, \text{hom}}(t, \mu_{\vec{x}}^N) = \mathbb{V}(t, \mu) = \bigcap_{\epsilon > 0} \limsup_{N \rightarrow \infty} \mathbb{V}_{\epsilon}^{N, \text{hom}}(t, \mu_{\vec{x}}^N)$$

Theorem

For $\mu_{\vec{x}}^N \Rightarrow \mu$,

$$\bigcap_{\epsilon > 0} \liminf_{N \rightarrow \infty} \mathbb{V}_{\epsilon}^N(t, \mu_{\vec{x}}^N) = \mathbb{V}^{\text{relax}}(t, \mu) = \bigcap_{\epsilon > 0} \limsup_{N \rightarrow \infty} \mathbb{V}_{\epsilon}^N(t, \mu_{\vec{x}}^N)$$

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N-player game

$$X_s^i = x_i + \int_t^s b(r, X_r^i, \mu_r^N, \alpha_i(r, X_r^i, \mu_r^N)) ds + B_s^i - B_t^i, \quad \mu_r^N = \frac{1}{N} \sum_j \delta_{X_r^j}$$

$$J_i(t, \vec{x}, (\alpha \tilde{\alpha}^i)) = J(t, \mu_t^N, \alpha; x_i, \tilde{\alpha}) = \mathbb{E} \left[G(X_T^i, \mu_T^N) + \int_t^T F(\dots) \right]$$

$$\alpha^* \in \mathcal{M}^\epsilon(t, \mu^N) : \int_{\mathbb{R}^d} [J(t, \mu^N, \alpha^*; x, \alpha^*) - \inf_{\tilde{\alpha}} J(t, \mu^N, \alpha^*; x, \tilde{\alpha})] \mu^N(dx) \leq \epsilon$$

Mean Field game

$$X_s^\alpha = \xi + \int_t^s b(r, X_r^\alpha, \mu_r, \alpha(r, X_r^\alpha, \mu_r)) ds + B_s - B_t, \quad \mu_r^\alpha = \mathcal{L}_{X_s^\alpha}$$

$$X_s^{\alpha, \tilde{\alpha}} = x + \int_t^s b(r, X_r^{\alpha, \tilde{\alpha}}, \mu_r^\alpha, \tilde{\alpha}(r, X_r^{\alpha, \tilde{\alpha}}, \mu_r^\alpha)) ds + B_s - B_t$$

$$J(t, \mu, \alpha; x, \tilde{\alpha}) = \mathbb{E} \left[G(X_T^{\alpha, \tilde{\alpha}}, \mu_T^\alpha) + \int_t^T F(\dots) \right]$$

$$\alpha^* \in \mathcal{M}^\epsilon(t, \mu) : \int_{\mathbb{R}^d} [J(t, \mu^N, \alpha^*; x, \alpha^*) - \inf_{\tilde{\alpha}} J(t, \mu, \alpha^*; x, \tilde{\alpha})] \mu(dx) \leq \epsilon$$

Main Results

We define the Set Value as

$$\mathbb{V}_\epsilon(t, \mu) \doteq \left\{ \varphi \in C_{Lip}(\mathbb{R}^d) : \exists \alpha^* \in \mathcal{M}^\epsilon(t, \mu) \text{ s.t. } \int_{\mathbb{R}^d} |\varphi(x) - J(t, \mu, \alpha^*; x, \alpha^*)| \mu(dx) \leq \epsilon \right\}$$

Theorem

$$\bigcap_{\epsilon > 0} \liminf_{N \rightarrow \infty} \mathbb{V}_\epsilon^{N, hom}(t, \mu_{\bar{x}}^N) = \mathbb{V}(t, \mu) = \bigcap_{\epsilon > 0} \limsup_{N \rightarrow \infty} \mathbb{V}_\epsilon^{N, hom}(t, \mu_{\bar{x}}^N)$$

Theorem

$$\mathbb{V}(t, \mu) = \bigcap_{\epsilon > 0} \left\{ \varphi \in C_{Lip}(\mathbb{R}^d) : \exists (\psi, \alpha^\epsilon) \text{ s.t.} \right. \\ \left. \alpha^\epsilon \in \mathcal{M}^\epsilon(T_0, \psi, t, \mu) \text{ and } \psi \in \mathbb{V}_\epsilon(T_0, \mu_{T_0}^{\alpha^\epsilon}) \right. \\ \left. \int_{\mathbb{R}^d} |\varphi(x) - J(T_0, \psi; t, \mu, \alpha^\epsilon; x, \alpha^\epsilon)| \mu(dx) \leq \epsilon \right\}$$

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