

Optimal control problems with generalized mean-field dynamics and viscosity solution to Master Bellman equation

Rainer Buckdahn

SDU, Qingdao, China, & UBO, Brest, France.

E-mail: rainer.buckdahn@univ-brest.fr

Based on a joint work with

Juan Li (Shandong University, Weihai & Qingdao, China)

Zhanxin Li (Shandong University, Weihai, China)

International H. & St. Geiß Seminar, 2025/06/06.

Outline

- 1 Objective of the talk
- 2 Formulation of the mean-field stochastic control problems
- 3 Dynamic programming principle
- 4 Master Bellman equation and viscosity solution
- 5 Main results

- 1 Objective of the talk
- 2 Formulation of the mean-field stochastic control problems
- 3 Dynamic programming principle
- 4 Master Bellman equation and viscosity solution
- 5 Main results

Brief state of the art

Mean-field problems:

- 1) Mean-Field SDEs have been intensively studied for a longer time as limit equ. for systems with a large number of particles (propagation of chaos)(Bossy, Méléard, Sznitman, Talay,...);
- 2) Mean-Field Games and related topics, since 2006-2007 by J.M.Lasry and P.L.Lions, Huang-Caines-Malhamé (2006) (Nash certainty equivalence principle); Mean field game systems:

$$\begin{cases} \text{i)} & -\partial_t u - v\Delta u + H(x, Du, m) = 0 \quad \text{in } (0, T) \times \mathbb{R}^d \text{ HJB equ.} \\ \text{ii)} & \partial_t m - v\Delta m - \operatorname{div}(H_p(x, Du, m)m) = 0 \quad \text{in } (0, T) \times \mathbb{R}^d \text{ continuity equ.} \\ \text{iii)} & m(0) = m_0, \quad u(x, T) = G(x, m(T)) \quad \text{in } \mathbb{R}^d \end{cases}$$

Master equation evaluated for $U = U(t, x, m)$:

$$\begin{cases} -\partial_t U - v\Delta_x U + H(x, D_x U, m) - v \int_{\mathbb{R}^d} \operatorname{div}_y D_m U(t, x, m, y) m(dy) \\ \quad + \int_{\mathbb{R}^d} D_m U(t, x, m, y) \cdot D_p H(y, D_x U, m) m(dy) = 0 \\ U(T, x, m) = G(x, m) \quad \text{in } \mathbb{R}^d \times \mathcal{P}_2(\mathbb{R}^d). \end{cases}$$

Brief state of the art

Mean-field problems:

3) +) Mean-Field BSDEs/FBSDEs and associated nonlocal PDEs:

- Prel. works: B., Djehiche, Li, Peng (AOP2009); B., Li, Peng (SPA2009);
- Classical solution of non-local PDE related with the mean-field SDE:
B., Li, Peng, Rainer (AOP2017 (Arxiv2014)):

$$\begin{aligned} 0 &= \partial_t V(t, x, \mu) + \partial_x V(t, x, \mu)b(x, \mu) + \frac{1}{2}\partial_{xx}^2 V(t, x, \mu)\sigma^2(x, \mu) \\ &\quad + \int_{\mathbb{R}} [(\partial_{\mu} V)(t, x, \mu, y)b(y, \mu) + \frac{1}{2}\partial_y(\partial_{\mu} V)(t, x, \mu, y)\sigma^2(y, \mu)]\mu(dy), \\ V(T, x, \mu) &= \Phi(x, \mu), \quad (t, x, \mu) \in [0, T] \times \mathbb{R} \times \mathcal{P}_2(\mathbb{R}). \end{aligned}$$

- For the case with mean-field SDE with jumps: Hao, Li (NODEA2016);
- For the case with the mean-field FBSDE with jumps: Li (SPA2017);
- For the case with the mean-field BDSDE and related nonlocal semi-linear backward SPDEs: B., Li, Xing (Arxiv2021);

Brief state of the art

Dynamic programming approach:

+) For classical case:

- For SDGs: B., Li (2008, SICON);
- For SDGs with jumps: B., Hu, Li (2011, SPA);

+) Classical case, but fully coupled FBSDEs with jumps: Li, Wei (2014, AMO);

- For stochastic control problems with jumps: Li, Peng (2009, NA);
- Stoch. control for fully coupled FBSDEs: Li, Wei (2014, SICON);

+) For mean-field stochastic optimal control problems:

The objective is to characterize the value function of the mean-field control problem as a viscosity solution of **a second order PDE on Wasserstein space**, known as **Master Bellman equation**. The viscosity theory of this kind of PDEs is still at a rather early stage.

- **Consider definition by “lifting” :**

The following works adopt the notion of viscosity solution from Crandall-Lions and adapt it to the Wasserstein space by lifting to L^2 . The uniqueness is established for this lifted Bellman equation.

↪ Pham, Wei (2018, ESAIM:COCV):

Controlled mean-field stochastic system:

$$dX_t^u = b(t, X_t^u, u_t, \mathbb{P}_{(X_t^u, u_t)})dt + \sigma(t, X_t^u, u_t, \mathbb{P}_{(X_t^u, u_t)})dW_t, \quad t \in [0, T],$$

where $u_t := \tilde{u}(t, X_t^u, \mathbb{P}_{X_t^u})$.

It is assumed a priori that the controls were of Markovian feedback type.

Brief state of the art

↪ Bayraktar, Cosso, and Pham (2018, TAMS): [Controlled mean-field SDEs](#):

$$X_s^{t,\xi,u} = \xi + \int_t^s b(r, X_r^{t,\xi,u}, \mathbb{P}_{X_r^{t,\xi,u}}, u_r) dr + \int_t^s \sigma(r, X_r^{t,\xi,u}, \mathbb{P}_{X_r^{t,\xi,u}}, u_r) dW_r,$$
$$X_s^{t,x,\xi,u} = x + \int_t^s b(r, X_r^{t,x,\xi,u}, \mathbb{P}_{X_r^{t,\xi,u}}, u_r) dr + \int_t^s \sigma(r, X_r^{t,x,\xi,u}, \mathbb{P}_{X_r^{t,\xi,u}}, u_r) dW_r.$$

They study control problems for **open-loop controls**, but without the dependence of the law on the control, and they proved a so-called Randomized DPP, based on a characterisation of the value function through an auxiliary intensity control problem for a Poisson random measure.

↪ Pham and Wei (2017, SICON),

↪ Cosso and Pham (2019, JMPA),

...

Brief state of the art

- **Consider “intrinsic” definition:**

↪ Burzoni, Ignazio, Reppen and Soner (2020, SICON):

Controlled mean -field stochastic system with jumps:

$$dX_t^u = b(t, \mathbb{P}_{X_t^u}, u_t)dt + \sigma(t, \mathbb{P}_{X_t^u}, u_t)dW_t + dJ_t, \quad t \in [0, T],$$

where J is a purely discontinuous process.

The authors considered deterministic control processes only depending on the time. They studied viscosity solutions for a particular class of integro-differential Master equations. The uniqueness of viscosity solutions has been proved on Wasserstein spaces of probability measures which have finite exponential moments.

↪ Cosso et al. (2024, TAMS): Controlled mean-field stochastic system:

$$dX_t^u = b(t, X_t^u, u_t, \mathbb{P}_{X_t^u})dt + \sigma(t, X_t^u, u_t)dW_t, \quad t \in [0, T].$$

By using refinements of early ideas from the Crandall-Lions theory of viscosity solutions, they proved the uniqueness of the viscosity solutions on $\mathcal{P}_2(\mathbb{R}^d)$, but only for coefficients which do not depend on the law of the control.

1. Objective of the talk

We develop a dynamic programming approach to study an optimal control problem with generalized mean-field dynamics with considering:

- Open-loop controls;
- Coefficients which depend on the joint law of state processes and controls;
- Dynamics of both a “mean-field player” and a representative “individual player”.

We characterize the value function as the unique viscosity solution of a second order PDE on Wasserstein space, by adapting the intrinsic notion of viscosity solutions in Burzoni et al. [2020].

- 1 Objective of the talk
- 2 Formulation of the mean-field stochastic control problems**
- 3 Dynamic programming principle
- 4 Master Bellman equation and viscosity solution
- 5 Main results

2. Formulation of the mean-field stochastic control problems

We consider:

- $(\Omega, \mathcal{F}, \mathbb{P})$ – the classical Wiener space; the driving Brownian Motion B is the coordinate process on $\Omega := C_0([0, T]; \mathbb{R}^d)$: $T > 0$ - a fixed horizon; $\mathcal{F} = \mathcal{B}(\Omega) \vee \mathcal{N}_{\mathbb{P}}$; \mathbb{P} - Wiener measure.
- $\mathbb{F} = \{\mathcal{F}_s, 0 \leq s \leq T\}$ – the filtration generated by $B = (B_s)_{s \in [0, T]}$ and augmented by all \mathbb{P} -null sets.
- For $k \geq 1$, $\mathcal{P}_2(\mathbb{R}^k)$ – the space of the probability measures over \mathbb{R}^k with finite second moment and endowed with the 2-Wasserstein metric:

$$\mathcal{W}_2(\mu, \nu) := \inf \left\{ \left(\int_{\mathbb{R}^k \times \mathbb{R}^k} |x - y|^2 \rho(dxdy) \right)^{\frac{1}{2}}, \rho \in \Pi_{\mu, \nu} \right\}, \mu, \nu \in \mathcal{P}_2(\mathbb{R}^k),$$

where $\Pi_{\mu, \nu} = \{\rho \in \mathcal{P}_2(\mathbb{R}^{2k}) \text{ with } \rho(\cdot \times \mathbb{R}^k) = \mu, \rho(\mathbb{R}^k \times \cdot) = \nu\}$.

Note: $(\mathcal{P}_2(\mathbb{R}^k), \mathcal{W}_2)$ is a complete separable space.

2. Formulation of the mean-field stochastic control problems

Spaces we work with:

- $L^2(\mathcal{F}_t; \mathbb{R}^n)$ is the set of \mathbb{R}^n -valued, \mathcal{F}_t -measurable random variables $\zeta : \Omega \rightarrow \mathbb{R}^n$ such that $E[|\zeta|^2] < \infty$.
- $L^2_{\mathbb{F}}([0, T]; \mathbb{R}^n)$ is the set of \mathbb{R}^n -valued, \mathbb{F} -progressively measurable processes $\phi : \Omega \times [0, T] \rightarrow \mathbb{R}^n$, with $E\left[\int_0^T |\phi_t|^2 dt\right] < +\infty$.
- $\mathcal{S}^2(0, T; \mathbb{R}^n)$ is the set of \mathbb{F} -adapted continuous processes $\phi : \Omega \times [0, T] \rightarrow \mathbb{R}^n$ satisfying $E\left[\sup_{0 \leq s \leq T} |\phi_s|^2\right] < \infty$.

For simplicity, we write $L^2(\mathcal{F}_t) := L^2(\mathcal{F}_t; \mathbb{R})$, $L^2_{\mathbb{F}}([0, T]) := L^2_{\mathbb{F}}([0, T]; \mathbb{R})$.

2. Formulation of the mean-field stochastic control problem

The **dynamics** of our stochastic control problem are the following controlled mean-field SDEs:

$$\begin{aligned} X_s^{t,\xi,u^2} = & \xi + \int_t^s b_1(r, (X_r^{t,\xi,u^2}, u_r^2), \mathbb{P}_{(X_r^{t,\xi,u^2}, u_r^2)}) dr \\ & + \int_t^s \sigma_1(r, (X_r^{t,\xi,u^2}, u_r^2), \mathbb{P}_{(X_r^{t,\xi,u^2}, u_r^2)}) dB_r, \quad s \in [t, T], \end{aligned} \quad (2.1)$$

$$\begin{aligned} X_s^{t,x,\xi,u} = & x + \int_t^s b_2(r, (X_r^{t,x,\xi,u}, u_r^1), \mathbb{P}_{(X_r^{t,\xi,u^2}, u_r^2)}) dr \\ & + \int_t^s \sigma_2(r, (X_r^{t,x,\xi,u}, u_r^1), \mathbb{P}_{(X_r^{t,\xi,u^2}, u_r^2)}) dB_r, \quad s \in [t, T], \end{aligned} \quad (2.2)$$

where $t \in [0, T]$, $x \in \mathbb{R}$, $\xi \in L^2(\mathcal{F}_t; \mathbb{R})$, $u = (u^1, u^2) \in \mathcal{U}_{t,T} := \mathcal{U}_{t,T}^0 \times \mathcal{U}_{t,T}^0$. Admissible controls $u^1, u^2 \in \mathcal{U}_{t,T}^0$ are \mathbb{F} -adapted stochastic processes on $[t, T]$ taking values in a compact set $U \subset \mathbb{R}^n$.

2. Formulation of the mean-field stochastic control problems

Remark: Interpretation of the dynamics

- Equation (2.1) can be interpreted as the dynamics of an agent (referred to as the “**mean-field player**”), who plays collectively with using a “collective control” u^2 . It describes **the average over the states of all agents**.
- Equation (2.2) describes the dynamics of **an individual agent who faces the “mean-field player”**, and u^1 is the “individual control” played by this individual agent.
- In other words, (2.1) characterizes the evolution of the law $\mathbb{P}_{(X^t, \zeta, u^2, u^2)}$, while (2.2) describes the associated trajectories of an individual agent with initial condition $X_t^{t, x, \zeta, u} = x$.

2. Formulation of the mean-field stochastic control problem

We shall make the following standard assumptions.

Assumption 2.1.

- (i) $b_1, b_2, \sigma_1, \sigma_2$ are continuous and, for simplicity, bounded.
- (ii) Lipschitz continuity: There exists a constant $C > 0$, such that

$$|\phi(t, (x, u), \mathbb{P}_{(\zeta, \eta)}) - \phi(t, (x', u), \mathbb{P}_{(\zeta', \eta)})| \leq C(|x - x'| + \mathcal{W}_2(\mathbb{P}_{(\zeta, \eta)}, \mathbb{P}_{(\zeta', \eta)})),$$

for all $t \in [0, T]$, $x, x' \in \mathbb{R}^n$, $u \in U$, $\zeta, \zeta' \in L^2(\mathcal{F}_t; \mathbb{R}^n)$, $\eta \in L^2(\mathcal{F}_t; U)$, $\phi = b_1, b_2, \sigma_1, \sigma_2$.

Under Assumption 2.1, there exists a unique pair of solutions $(X_s^{t, \zeta, u^2}, X_s^{t, x, \zeta, u})_{s \in [t, T]} \in \mathcal{S}^2(0, T; \mathbb{R}^n) \times \mathcal{S}^2(0, T; \mathbb{R}^n)$ to the equations (2.1) and (2.2) (see, e.g., Buckdahn, Li, Peng and Rainer [2017]).

2. Formulation of the mean-field stochastic control problem

Moreover, for every $p \geq 2$, we have the following L^p -**estimates**:
There exists $C_p \in \mathbb{R}^+$ such that, for all $t \in [0, T]$, $u = (u^1, u^2) \in \mathcal{U}_{t,T}$,
and $x, x' \in \mathbb{R}^n$, $\zeta, \zeta' \in L^2(\mathcal{F}_t; \mathbb{R}^n)$,

$$E \left[\sup_{s \in [t, T]} \left| X_s^{t, \zeta, u^2} - X_s^{t, \zeta', u^2} \right|^p \middle| \mathcal{F}_t \right] \leq C_p |\zeta - \zeta'|^p,$$

$$E \left[\sup_{s \in [t, T]} \left| X_s^{t, x, \zeta, u} - X_s^{t, x', \zeta', u} \right|^p \middle| \mathcal{F}_t \right] \leq C_p \left(|x - x'|^p + |\zeta - \zeta'|^p \right).$$

From the uniqueness of the solutions of the both equations, we also have the following **flow property**: For all $0 \leq t < t + \delta \leq T$, $x \in \mathbb{R}^n$, $\zeta \in L^2(\mathcal{F}_t; \mathbb{R}^n)$, $u = (u^1, u^2) \in \mathcal{U}_{t,T}$:

$$\left(X_s^{t+\delta, X_{t+\delta}^{t, x, \zeta, u}, X_{t+\delta}^{t, \zeta, u^2}, u}, X_s^{t+\delta, X_{t+\delta}^{t, \zeta, u^2}, u^2} \right) = \left(X_s^{t, x, \zeta, u}, X_s^{t, \zeta, u^2} \right), s \in [t+\delta, T], \mathbb{P}\text{-a.s.}$$

2. Formulation of the mean-field stochastic control problem

Assumption 2.2.

Let $\Phi : \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathbb{R}$ be Lipschitz, i.e., for some constant $C > 0$ we have, for all $x, x' \in \mathbb{R}^n, \mu, \mu' \in \mathcal{P}_2(\mathbb{R}^n)$,

$$|\Phi(x, \mu) - \Phi(x', \mu')| \leq C(|x - x'| + \mathcal{W}_2(\mu, \mu')).$$

Given the control processes $u = (u^1, u^2) \in \mathcal{U}_{t,T}$, we introduce the **cost functional** of our mean-field stochastic control problem:

$$J(t, x, \zeta, u) := E[\Phi(X_T^{t,x,\zeta,u}, \mathbb{P}_{X_T^{t,\zeta,u^2}}) | \mathcal{F}_t],$$

where $(t, x) \in [0, T] \times \mathbb{R}^n, \zeta \in L^2(\mathcal{F}_t; \mathbb{R}^n), (X^{t,\zeta,u^2}, X^{t,x,\zeta,u})$ are the solutions of (2.1) and (2.2).

Remark. $X^{t,x,\zeta,u}$ is in general not independent of \mathcal{F}_t , and therefore $J(t, x, \zeta, u)$ is an \mathcal{F}_t -measurable random variable.

2. Formulation of the mean-field stochastic control problem

The definition of the value function:

Suppose that both the “mean-field player” and the “individual player” try to minimize the cost, for all $(t, x) \in [0, T] \times \mathbb{R}^n$, $\zeta \in L^2(\mathcal{F}_t; \mathbb{R}^n)$,

$$\begin{aligned} V(t, x, \zeta) &:= \operatorname{essinf}_{u \in \mathcal{U}_{t,T}} J(t, x, \zeta, u) \\ &= \operatorname{essinf}_{u^2 \in \mathcal{U}_{t,T}^0} \left(\operatorname{essinf}_{u^1 \in \mathcal{U}_{t,T}^0} J(t, x, \zeta, (u^1, u^2)) \right). \end{aligned}$$

Set $W(t, x, \zeta, u^2) := \operatorname{essinf}_{u^1 \in \mathcal{U}_{t,T}^0} J(t, x, \zeta, (u^1, u^2))$, $(t, x) \in [0, T] \times \mathbb{R}^n$, $u^2 \in \mathcal{U}_{t,T}^0$. Then

$$V(t, x, \zeta) = \operatorname{essinf}_{u^2 \in \mathcal{U}_{t,T}^0} W(t, x, \zeta, u^2).$$

- 1 Objective of the talk
- 2 Formulation of the mean-field stochastic control problems
- 3 Dynamic programming principle**
- 4 Master Bellman equation and viscosity solution
- 5 Main results

3. Dynamic programming principle

Recall that $W(t, x, \zeta, u^2)$ is an \mathcal{F}_t -measurable random variable. The following lemma shows that it is even deterministic.

Lemma 3.1.

For any $(t, x) \in [0, T] \times \mathbb{R}^n$, $\zeta \in L^2(\mathcal{F}_t; \mathbb{R}^n)$, $u^2 \in \mathcal{U}_{t,T}^0$, $W(t, x, \zeta, u^2)$ is **deterministic**. Moreover, by standard estimates, $\exists C > 0$ s.t.

$$|W(t, x, \zeta, u^2) - W(t, x', \zeta', u^2)| \leq C(|x - x'| + (E[|\zeta - \zeta'|^2])^{\frac{1}{2}}),$$

for all $t \in [0, T]$, $x, x' \in \mathbb{R}^n$, $\zeta, \zeta' \in L^2(\mathcal{F}_t; \mathbb{R}^n)$, $u^2 \in \mathcal{U}_{t,T}^0$.

The proof of this lemma uses a standard argument from Buckdahn and Li [2008] with a novel, subtle approach.

The above lemma, combined with the definition of the value function V yields that also

$$V(t, x, \zeta) = \inf_{u^2 \in \mathcal{U}_{t,T}^0} W(t, x, \zeta, u^2)$$

is deterministic, for all $(t, x) \in [0, T] \times \mathbb{R}^n$, $\zeta \in L^2(\mathcal{F}_t; \mathbb{R}^n)$.

3. Dynamic programming principle

Next, we prove that **the value function** $V(t, x, \zeta)$ **does not depend on ζ itself but only on its law** \mathbb{P}_ζ . For this, we have first to study some auxiliary results.

Consider the following space of elementary control processes:

$$\mathcal{U}_{t,T}^e := \left\{ u^2 = \sum_{i,j=0}^{N-1} \mathbf{1}_{A_{i,j}} \zeta_{i,j} \mathbf{1}_{(t_i, t_{i+1}]} \mid N \geq 1, t = t_0 \leq \dots \leq t_N = T, A_{i,j} \in \mathcal{F}_t, \right. \\ \left. \zeta_{i,j} \in L^2(\mathcal{F}_{t_i}^t; U), i = 0, \dots, N-1, (A_{i,j})_{j=0}^{N-1} \text{ is a decomposition of } \Omega \right\},$$

where $\mathcal{F}_s^t := \sigma\{B_r - B_t, r \in [t, s]\}$, $s \in [t, T]$.

Lemma 3.2.

$\mathcal{U}_{t,T}^e$ is dense in $L_{\mathbb{F}}^2([t, T]; U) (= \mathcal{U}_{t,T}^0)$ with respect to the L^2 -norm over $[t, T] \times \Omega$.

3. Dynamic programming principle

Then standard arguments allow to show:

$$V(t, x, \zeta) = \inf_{u^2 \in \mathcal{U}_{t,T}^e} W(t, x, \zeta, u^2).$$

Now we have the following result:

Lemma 3.4.

For every $\zeta, \zeta' \in L^2(\mathcal{F}_t; \mathbb{R}^n)$ with $\mathbb{P}_\zeta = \mathbb{P}_{\zeta'}$, and every $u^2 \in \mathcal{U}_{t,T}^e$, there exists $u^{2'} \in \mathcal{U}_{t,T}^e$ such that

$$J(t, x, \zeta', (u^1, u^{2'})) = J(t, x, \zeta, (u^1, u^2)), \quad \mathbb{P}\text{-a.s.},$$

for all $u^1 \in \mathcal{U}_{t,T}^0$, $x \in \mathbb{R}^n$, and, in particular,

$$\begin{aligned} W(t, x, \zeta', u^{2'}) &:= \operatorname{essinf}_{u^1 \in \mathcal{U}_{t,T}^0} J(t, x, \zeta', (u^1, u^{2'})) \\ &= \operatorname{essinf}_{u^1 \in \mathcal{U}_{t,T}^0} J(t, x, \zeta, (u^1, u^2)) = W(t, x, \zeta, u^2). \end{aligned}$$

3. Dynamic programming principle

From the above lemma we deduce now easily:

Proposition 3.2.

Let $(t, x) \in [0, T] \times \mathbb{R}^n$, $\zeta \in L^2(\mathcal{F}_t; \mathbb{R}^n)$. Then, for all $\zeta' \in L^2(\mathcal{F}_t; \mathbb{R}^n)$ with $\mathbb{P}_\zeta = \mathbb{P}_{\zeta'}$, we have

$$V(t, x, \zeta) = V(t, x, \zeta'),$$

i.e., V depends on ζ only through \mathbb{P}_ζ . We write:

$$V(t, x, \mathbb{P}_\zeta) := V(t, x, \zeta),$$

where $V : [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathbb{R}$. Moreover, by standard estimates, there exists $C \in \mathbb{R}_+$, such that

$$|V(t, x, \mathbb{P}_\zeta) - V(t, \tilde{x}, \mathbb{P}_{\tilde{\zeta}})| \leq C(|x - \tilde{x}| + \mathcal{W}_2(\mathbb{P}_\zeta, \mathbb{P}_{\tilde{\zeta}})),$$

for all $t \in [0, T]$, $x, \tilde{x} \in \mathbb{R}^n$, $\zeta, \tilde{\zeta} \in L^2(\mathcal{F}_t; \mathbb{R}^n)$.

3. Dynamic programming principle

Study of $W(t, x, \zeta, u^2)$:

Recall that:

- $W(t, \cdot, \cdot, \cdot) : \mathbb{R} \times L^2(\mathcal{F}_t; \mathbb{R}^n) \times \mathcal{U}_{t,T}^0 \rightarrow \mathbb{R}$ is defined by
$$W(t, x, \zeta, u^2) = \operatorname{essinf}_{u^1 \in \mathcal{U}_{t,T}^0} J(t, x, \zeta, (u^1, u^2))$$
$$= \operatorname{essinf}_{u^1 \in \mathcal{U}_{t,T}^0} E[\Phi(X_T^{t,x,\zeta,(u^1,u^2)}, \mathbb{P}_{X_T^{t,\zeta,u^2}}) | \mathcal{F}_t];$$
- For any $u^2 \in \mathcal{U}_{t,T}^0$, $W(t, x, \zeta, u^2)$ is deterministic;
- From standard estimates:

$$|W(t, x, \zeta, u^2) - W(t, x', \zeta', u^2)| \leq C \left(|x - x'| + (E[|\zeta - \zeta'|^2])^{\frac{1}{2}} \right),$$

for all $t \in [0, T]$, $x, x' \in \mathbb{R}^n$, $\zeta, \zeta' \in L^2(\mathcal{F}_t; \mathbb{R}^n)$, $u^2 \in \mathcal{U}_{t,T}^0$.

3. Dynamic programming principle

Using the above properties of W , we obtain the following DPP for W .

Theorem 3.1. (DPP for W)

For all $0 \leq t < t + \delta \leq T$, $u^2 \in \mathcal{U}_{t,T}^0$, $x \in \mathbb{R}^n$, $\zeta \in L^2(\mathcal{F}_t; \mathbb{R}^n)$,

$$W(t, x, \zeta, u^2) = \operatorname{ess\,inf}_{u^1 \in \mathcal{U}_{t,t+\delta}^0} E[W(t + \delta, X_{t+\delta}^{t,x,\zeta,(u^1,u^2)}, X_{t+\delta}^{t,\zeta,u^2}, u^2) | \mathcal{F}_t].$$

The proof of this proposition uses a standard argument from Theorem 3.1 of Buckdahn and Li [2008].

3. Dynamic programming principle

Remark.

From the proof of Theorem 3.1, as W is deterministic, we obtain, for all $t < t + \delta \leq T$,

$$W(t, x, \zeta, u^2) = \inf_{u^1 \in \mathcal{U}_{t, t+\delta}^0} E \left[W(t + \delta, X_{t+\delta}^{t, x, \zeta, (u^1, u^2)}, X_{t+\delta}^{t, \zeta, u^2}, u^2) \right]. \quad (3.1)$$

In particular, for $t + \delta = T$,

$$W(t, x, \zeta, u^2) = \inf_{u^1 \in \mathcal{U}_{t, T}^0} E \left[\Phi(X_T^{t, x, \zeta, (u^1, u^2)}, \mathbb{P}_{X_T^{t, \zeta, u^2}}) \right],$$

and therefore, for the value function $V(t, x, \mathbb{P}_\zeta) = V(t, x, \zeta)$:

$$\begin{aligned} V(t, x, \mathbb{P}_\zeta) &= V(t, x, \zeta) = \operatorname{ess\,inf}_{(u^1, u^2) \in \mathcal{U}_{t, T}} E \left[\Phi(X_T^{t, x, \zeta, (u^1, u^2)}, \mathbb{P}_{X_T^{t, \zeta, u^2}}) \middle| \mathcal{F}_t \right] \\ &= \inf_{u^2 \in \mathcal{U}_{t, T}^0} W(t, x, \zeta, u^2) = \inf_{(u^1, u^2) \in \mathcal{U}_{t, T}} E \left[\Phi(X_T^{t, x, \zeta, (u^1, u^2)}, \mathbb{P}_{X_T^{t, \zeta, u^2}}) \right]. \end{aligned}$$

3. Dynamic programming principle

Study of $V(t, x, \mathbb{P}_\zeta)$:

As the following inequality shows, **we get the one-sided DPP for V :**
For $0 \leq t < t + \delta \leq T$, $x \in \mathbb{R}^n$, $\zeta \in L^2(\mathcal{F}_t; \mathbb{R}^n)$,

$$\begin{aligned} V(t, x, \mathbb{P}_\zeta) &\stackrel{V \text{ Def}}{=} \inf_{u^2 \in \mathcal{U}_{t,T}^0} W(t, x, \zeta, u^2) \\ &\stackrel{(3.1)}{=} \inf_{u^2 \in \mathcal{U}_{t,T}^0} \left(\inf_{u^1 \in \mathcal{U}_{t,t+\delta}^0} E[W(t + \delta, X_{t+\delta}^{t,x,\zeta,(u^1,u^2)}, X_{t+\delta}^{t,\zeta,u^2}, u^2)] \right) \\ &\stackrel{W \text{ Def}}{\geq} \inf_{u^2 \in \mathcal{U}_{t,T}^0} \left(\inf_{u^1 \in \mathcal{U}_{t,t+\delta}^0} E[V(t + \delta, X_{t+\delta}^{t,x,\zeta,(u^1,u^2)}, X_{t+\delta}^{t,\zeta,u^2})] \right) \\ &\stackrel{\text{Indep. of } u^2|_{(t+\delta,T)}}{=} \inf_{(u^1, u^2) \in \mathcal{U}_{t,t+\delta}} E[V(t + \delta, X_{t+\delta}^{t,x,\zeta,(u^1,u^2)}, \mathbb{P}_{X_{t+\delta}^{t,\zeta,u^2}})]. \end{aligned}$$

But we can not get the above inequality in the opposite direction.

3. Dynamic programming principle

The value function ϑ :

For $\theta, \zeta \in L^2(\mathcal{F}_t; \mathbb{R}^n)$, we introduce a new definition of the value function:

$$\begin{aligned}\vartheta(t, \theta, \mathbb{P}_\zeta) &:= \inf_{u^2 \in \mathcal{U}_{t,T}^0} E[W(t, \theta, \zeta, u^2)] \\ &= \inf_{u^2 \in \mathcal{U}_{t,T}^0} E \left[\operatorname{ess\,inf}_{u^1 \in \mathcal{U}_{t,T}^0} E[\Phi(X_T^{t,\theta,\zeta,(u^1,u^2)}, \mathbb{P}_{X_T^{t,\zeta,u^2}}) | \mathcal{F}_t] \right].\end{aligned}\tag{3.2}$$

We can see:

- The function ϑ is obviously deterministic;
- $\vartheta(t, \theta, \mathbb{P}_\zeta)$ depends on θ only through the law \mathbb{P}_θ . (Indeed,

$$\vartheta(t, \theta, \mathbb{P}_\zeta) = \inf_{u^2 \in \mathcal{U}_{t,T}^0} \int_{\mathbb{R}^n} W(t, x, \zeta, u^2) \mathbb{P}_\theta(dx).$$

This allows to write

$$\vartheta(t, \mathbb{P}_\theta, \mathbb{P}_\zeta) := \vartheta(t, \theta, \mathbb{P}_\zeta),$$

and to consider ϑ as a function over $[0, T] \times \mathcal{P}_2(\mathbb{R}^n) \times \mathcal{P}_2(\mathbb{R}^n)$.

3. Dynamic programming principle

From Lemma 3.1 and a standard argument we have the following estimate for ϑ : There exists a constant $C > 0$ such that

$$|\vartheta(t, \mathbb{P}_\theta, \mathbb{P}_\zeta) - \vartheta(t, \mathbb{P}_{\theta'}, \mathbb{P}_{\zeta'})| \leq C(\mathcal{W}_2(\mathbb{P}_\theta, \mathbb{P}_{\theta'}) + \mathcal{W}_2(\mathbb{P}_\zeta, \mathbb{P}_{\zeta'})), \quad (3.3)$$

for all $t \in [0, T]$, $\theta, \theta', \zeta, \zeta' \in L^2(\mathcal{F}_t; \mathbb{R}^n)$.

Remark.

Notice that, for $(t, x) \in [0, T] \times \mathbb{R}^n$, $\zeta \in L^2(\mathcal{F}_t; \mathbb{R}^n)$,

$$\vartheta(t, x, \mathbb{P}_\zeta)(:= \vartheta(t, \delta_x, \mathbb{P}_\zeta)) = V(t, x, \mathbb{P}_\zeta),$$

where δ_x denotes the Dirac measure at x , which means that **a description of $\vartheta(t, \mathbb{P}_\theta, \mathbb{P}_\zeta)$, $(\theta, \zeta) \in L^2(\mathcal{F}_t; \mathbb{R}^n) \times L^2(\mathcal{F}_t; \mathbb{R}^n)$, as a solution of a PDE also characterizes $V(t, x, \mathbb{P}_\zeta)$.** This allows to characterize V through studying ϑ .

3. Dynamic programming principle

Let us now study the new value function ϑ . To begin with, we prove that ϑ obeys the following DPP:

Theorem 3.2. (DPP for ϑ)

For any $0 \leq t < t + \delta \leq T$, $\theta, \zeta \in L^2(\mathcal{F}_t; \mathbb{R}^n)$,

$$\vartheta(t, \mathbb{P}_\theta, \mathbb{P}_\zeta) = \inf_{u \in \mathcal{U}_{t, t+\delta}} \vartheta(t + \delta, \mathbb{P}_{X_{t+\delta}^{t, \theta, \zeta, u}}, \mathbb{P}_{X_{t+\delta}^{t, \zeta, u^2}}). \quad (3.4)$$

By using the properties of W , especially the DPP for W , and standard arguments from Buckdahn and Li [2008], we prove Theorem 3.2.

3. Dynamic programming principle

Using the continuity properties of $\vartheta(t, \cdot, \cdot)$ on $\mathcal{P}^2(\mathbb{R}^d) \times \mathcal{P}^2(\mathbb{R}^d)$ and the DPP for ϑ , we also prove the continuity of ϑ with respect to t .

Proposition 3.3.

The value function ϑ is $\frac{1}{2}$ -Hölder continuous in t : There exists a constant C such that, for every $t, t' \in [0, T]$, $\theta, \zeta \in L^2(\mathcal{F}_t; \mathbb{R}^n)$,

$$|\vartheta(t, \mathbb{P}_\theta, \mathbb{P}_\zeta) - \vartheta(t', \mathbb{P}_\theta, \mathbb{P}_\zeta)| \leq C(1 + (E[|\zeta|^2])^{\frac{1}{2}} + (E[|\theta|^2])^{\frac{1}{2}}) |t - t'|^{\frac{1}{2}}.$$

- 1 Objective of the talk
- 2 Formulation of the mean-field stochastic control problems
- 3 Dynamic programming principle
- 4 Master Bellman equation and viscosity solution**
- 5 Main results

4. Master Bellman equation and viscosity solution

Let us begin with the notion of derivative w.r.t. the measure over the Wasserstein space.

Definition 4.1. (Carmona and Delarue [2018])

A function $\varphi : \mathcal{P}_2(\mathbb{R}^k) \rightarrow \mathbb{R}$ is said to have a **linear functional derivative** if there exists a function

$$\frac{\delta\varphi}{\delta\mu} : \mathcal{P}_2(\mathbb{R}^k) \times \mathbb{R}^k \ni (\mu, x) \mapsto \frac{\delta\varphi}{\delta\mu}(\mu)(x) \in \mathbb{R},$$

which is continuous with respect to the product topology ($\mathcal{P}_2(\mathbb{R}^k)$ is equipped with the 2-Wasserstein distance) such that, for any bounded subset $\mathcal{K} \subset \mathcal{P}_2(\mathbb{R}^k)$, the function $\mathbb{R}^d \ni x \mapsto [\delta\varphi/\delta\mu](\mu)(x)$ is at most of quadratic growth in x , uniformly in μ , for $\mu \in \mathcal{K}$, and for all μ and μ' in $\mathcal{P}_2(\mathbb{R}^k)$, it holds:

$$\varphi(\mu') - \varphi(\mu) = \int_0^1 \int_{\mathbb{R}^d} \frac{\delta\varphi}{\delta\mu}(r\mu' + (1-r)\mu)(x) d(\mu' - \mu)(x) dr.$$

4. Master Bellman equation and viscosity solution

This leads to the definition of the L-derivative of φ (the derivative introduced by P.L. Lions [2013]).

Definition 4.2. (Carmona and Delarue [2018])

If $\frac{\delta\varphi}{\delta\mu}$ is of class \mathcal{C}^1 with respect to the second variable, the **L-derivative** $\partial_\mu\varphi : \mathcal{P}(\mathbb{R}^k) \times \mathbb{R}^k \rightarrow \mathbb{R}^k$ is defined by

$$\partial_\mu\varphi(\mu, x) := \partial_x\left(\frac{\delta\varphi}{\delta\mu}\right)(\mu, x), \quad (\mu, x) \in \mathcal{P}_2(\mathbb{R}^k) \times \mathbb{R}^k.$$

For any $\mu \in \mathcal{P}_2(\mathbb{R}^k)$ and any μ -integrable function $f : \mathbb{R}^k \rightarrow \mathbb{R}$, we use the notation

$$\langle \mu, f \rangle := \int_{\mathbb{R}^k} f(x) \mu(dx).$$

For simplicity of notation, we restrict ourselves to dimension $n = d = k = 1$ in what follows.

4. Master Bellman equation and viscosity solution

In this section, we study our optimal control problem on the space $\mathcal{O} := [0, T) \times \mathcal{M} \times \mathcal{M}$, where $\mathcal{M} \subset \mathcal{P}_2(\mathbb{R})$ is the set of probability measures with δ -exponential moments, i.e.,

$$\mathcal{M} := \left\{ \mu \in \mathcal{P}_2(\mathbb{R}) : \langle \mu, \exp(\delta|\cdot|) \rangle = \int_{\mathbb{R}} \exp(\delta|x|) \mu(dx) < \infty \right\}, \quad (4.1)$$

where $\delta > 0$ is an arbitrary given constant. We endow \mathcal{M} with the topology of weak convergence $\sigma(\mathcal{M}, C_b(\mathbb{R}))$, where $C_b(\mathbb{R})$ is the space of continuous and bounded functions on \mathbb{R} .

Remark. The space \mathcal{O} has a suitable σ -**compact structure**, which allows to establish uniqueness in the next section. This form of \mathcal{O} is crucial to obtain uniform integrability of the viscosity test functions as well as some continuity properties of the Hamiltonian.

4. Master Bellman equation and viscosity solution

Master Bellman equation:

For any $(t, \mu_1, \mu_2) \in [0, T) \times \mathcal{M} \times \mathcal{M}$,

$$\begin{cases} -\partial_t \vartheta(t, \mu_1, \mu_2) - \mathcal{H}(t, \mu_1, \mu_2, \partial_{\mu_1} \vartheta(t, \mu_1, \mu_2; \cdot), \partial_{\mu_2} \vartheta(t, \mu_1, \mu_2; \cdot)) = 0, \\ \vartheta(T, \mu_1, \mu_2) = \langle \mu_1, \Phi(\cdot, \mu_2) \rangle, \end{cases} \quad (4.2)$$

where, with the notation

$$\Pi_\mu := \{\gamma \in \mathcal{P}_2(\mathbb{R} \times U) : \gamma(\cdot \times U) = \mu\}, \quad \mu \in \mathcal{P}_2(\mathbb{R}),$$

the Hamiltonian \mathcal{H} is defined by

$$\begin{aligned} & \mathcal{H}(t, \mu_1, \mu_2, p_1, p_2) \\ & := \inf \left\{ \langle \gamma_1, \mathcal{L}_t^{\mu_1, \mu_2, \gamma_2}[p_1] \rangle + \langle \gamma_2, \bar{\mathcal{L}}_t^{\mu_1, \mu_2, \gamma_2}[p_2] \rangle : \gamma_i \in \Pi_{\mu_i}, i = 1, 2 \right\}, \end{aligned}$$

$p_1, p_2 \in C^1(\mathbb{R})$, with, for $(y, v) \in \mathbb{R} \times U$,

$$\mathcal{L}_t^{\mu_1, \mu_2, \gamma_2}[p_1](y, v) := p_1(y) b_2(t, (y, v), \gamma_2) + \frac{1}{2} \partial_y p_1(y) (\sigma_2(t, (y, v), \gamma_2))^2,$$

$$\bar{\mathcal{L}}_t^{\mu_1, \mu_2, \gamma_2}[p_2](y, v) := p_2(y) b_1(t, (y, v), \gamma_2) + \frac{1}{2} \partial_y p_2(y) (\sigma_1(t, (y, v), \gamma_2))^2.$$

4. Master Bellman equation and viscosity solution

Remark.

When the coefficients b_1 and σ_1 do not depend on the control u^2 , then $\vartheta(t, \delta_x, \mu) = V(t, x, \mu) = W(t, x, \mu)$, and from PDE (4.2) we get the following PDE related with $V(t, x, \mu)$: For $(t, x, \mu) \in [0, T) \times \mathbb{R} \times \mathcal{M}$,

$$\begin{cases} -\partial_t V(t, x, \mu) - \inf_{u \in U} \left\{ \partial_x V(t, x, \mu) b_2(t, (x, u), \mu) + \frac{1}{2} \partial_x^2 V(t, x, \mu) (\sigma_2(t, (x, u), \mu))^2 \right\} \\ - \int_{\mathbb{R}} \partial_\mu V(t, x, \mu; y) b_1(t, y, \mu) \mu(dy) - \int_{\mathbb{R}} \frac{1}{2} \partial_y \partial_\mu V(t, x, \mu; y) (\sigma_1(t, y, \mu))^2 \mu(dy) = 0, \\ V(T, x, \mu) = \Phi(x, \mu). \end{cases} \quad (4.3)$$

- If $b_1 = \sigma_1 = 0$, and b_2, σ_2, Φ do not depend on the law, PDE (4.3) is just the classical HJB equation.
- If $b_1 = b_2$, and $\sigma_1 = \sigma_2$, i.e., all the coefficients are free of controls, then PDE (4.2) is just the mean-field PDE obtained in Buckdahn, Li, Peng and Rainer [2017].

4. Master Bellman equation and viscosity solution

For δ as in (4.1), we consider the function

$$e_\delta(x) := \exp(\delta(\sqrt{x^2 + 1} - 1)), \quad x \in \mathbb{R}.$$

For $N \in \mathbb{N}$ and δ as in (4.1), let

$$\mathcal{O}_N := \left\{ (t, \mu_1, \mu_2) \in [0, T) \times \mathcal{P}_2(\mathbb{R}) \times \mathcal{P}_2(\mathbb{R}) \mid \langle \mu_i, e_\delta \rangle \leq N e^{K^* t}, i = 1, 2 \right\},$$

$$\overline{\mathcal{O}}_N := \left\{ (t, \mu_1, \mu_2) \in [0, T] \times \mathcal{P}_2(\mathbb{R}) \times \mathcal{P}_2(\mathbb{R}) \mid \langle \mu_i, e_\delta \rangle \leq N e^{K^* t}, i = 1, 2 \right\},$$

where K^* is a given positive constant which is derived from the proof of Lemma 4.1 further down, in order to ensure that \mathcal{O}_N is invariant for the dynamics (2.1)-(2.2). Note that $\mathcal{O} = \cup_{N=1}^\infty \mathcal{O}_N$ and $\overline{\mathcal{O}} = \cup_{N=1}^\infty \overline{\mathcal{O}}_N$.

For a constant b and δ as in (4.1), we put

$$\mathcal{M}_b := \{ \mu \in \mathcal{P}_2(\mathbb{R}) \mid \langle \mu, e_\delta \rangle \leq b \}.$$

4. Master Bellman equation and viscosity solution

Lemma 4.1.

Under Assumption 2.1 , for all $N \in \mathbb{N}$, the set \mathcal{O}_N is invariant for the SDEs (2.1)-(2.2), namely,

$$(t, \mathbb{P}_\theta, \mathbb{P}_\zeta) \in \mathcal{O}_N \implies \left(s, \mathbb{P}_{X_s^{t,\theta,\zeta,u}}, \mathbb{P}_{X_s^{t,\zeta,u^2}} \right) \in \mathcal{O}_N,$$

for all $t \in [0, T]$, $s \in [t, T]$, $\theta, \zeta \in L^2(\mathcal{F}_t)$, and $u = (u^1, u^2) \in \mathcal{U}_{t,T}$, where $(X_s^{t,\zeta,u^2}, X_s^{t,\theta,\zeta,u})_{s \in [t,T]}$ is the solution to (2.1)-(2.2) with initial condition $(X_t^{t,\zeta,u^2}, X_t^{t,\theta,\zeta,u}) = (\zeta, \theta)$.

This implies that for any given initial law $(t, \mu_1, \mu_2) \in \mathcal{O}_N$, we may restrict the Master Bellman equation (4.2) to \mathcal{O}_N .

4. Master Bellman equation and viscosity solution

As $\mathcal{P}_2(\mathbb{R})$ itself is not σ -compact, the importance of $\overline{\mathcal{O}}_N$ stems also from the following fact:

Lemma 4.2.

For $N \in \mathbb{N}$, $\overline{\mathcal{O}}_N$ is a compact subset of $[0, T] \times \mathcal{P}_2(\mathbb{R}) \times \mathcal{P}_2(\mathbb{R})$.

Now we give **the definition of a test function** and **that of a viscosity solution to the Master Bellman equation (4.2)**, which were first introduced in Burzoni et al. [2020]. We adapt them here to our framework.

Definition 4.3.

A **cylindrical function** is a mapping of the form $(t, \mu_1, \mu_2) \mapsto F(t, \langle \mu_1, f_1 \rangle, \langle \mu_2, f_2 \rangle)$ for some functions $f_1, f_2 : \mathbb{R} \rightarrow \mathbb{R}$ and $F : [0, T] \times \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$. This function is called **cylindrical polynomial**, if f_1, f_2 are polynomials and F is continuously differentiable.

4. Master Bellman equation and viscosity solution

We extend the above class to its linear span. For any polynomial f , we denote the degree of f by $\deg(f)$.

Definition 4.4. (Test Functions)

For $E \subset \overline{\mathcal{O}}$, a **viscosity test function** on E is a function of the form

$$\varphi(t, \mu_1, \mu_2) = \sum_{j=1}^{\infty} \varphi_j(t, \mu_1, \mu_2), \quad (t, \mu_1, \mu_2) \in E,$$

where $\{\varphi_j\}_{j \in \mathbb{N}}$ is a sequence of cylindrical polynomials which is absolutely convergent at every (t, μ_1, μ_2) and, for $i = 1, 2$,

$$\lim_{M \rightarrow \infty} \sum_{j=M}^{\infty} \sup_{(t, \mu_1, \mu_2) \in \mathcal{O}_N} \sum_{i=1,2} \left(\langle \mu_i, |\partial_{\mu_i} \varphi_j(t, \mu_1, \mu_2; \cdot)| \rangle + \langle \mu_i, |\partial_y \partial_{\mu_i} \varphi_j(t, \mu_1, \mu_2; \cdot)| \rangle \right) = 0. \quad (4.4)$$

Let Γ_E be the set of all viscosity test functions on E .

Remark. Condition (4.4) is used for proving the continuity of $(t, \mu_1, \mu_2) \mapsto \mathcal{H}(t, \mu_1, \mu_2, \partial_{\mu_1} \varphi(t, \mu_1, \mu_2; \cdot), \partial_{\mu_2} \varphi(t, \mu_1, \mu_2; \cdot))$ in Proposition 4.1, when $\varphi \in \Gamma_{\mathcal{O}_N}$.

4. Master Bellman equation and viscosity solution

Definition 4.5.

For $E \subseteq \overline{\mathcal{O}}$ and $(t, \mu_1, \mu_2) \in E$ with $t < T$, the **superjet** of a function $u : E \rightarrow \mathbb{R}$ at (t, μ_1, μ_2) is given by

$$J_E^+ u(t, \mu_1, \mu_2) := \left\{ (\partial_t \varphi(t, \mu_1, \mu_2), \partial_{\mu_1} \varphi(t, \mu_1, \mu_2; \cdot), \partial_{\mu_2} \varphi(t, \mu_1, \mu_2; \cdot)) \mid \right. \\ \left. \varphi \in \Gamma_E, (u - \varphi)(t, \mu_1, \mu_2) = \max_E (u - \varphi) \right\}.$$

The **subjet** of u at (t, μ_1, μ_2) is defined as

$$J_E^- u(t, \mu_1, \mu_2) := -J_E^+ (-u(t, \mu_1, \mu_2)).$$

4. Master Bellman equation and viscosity solution

Definition 4.6. (Viscosity Solution)

A continuous function $u : \mathcal{O}_N \rightarrow \mathbb{R}$ with $u(T, \mu_1, \mu_2) = \langle \mu_1, \Phi(\cdot, \mu_2) \rangle$ is called a **viscosity subsolution** of (4.2) on \mathcal{O}_N if, for every $(t, \mu_1, \mu_2) \in \mathcal{O}_N$,

$$-\pi_t - \mathcal{H}(t, \mu_1, \mu_2, \pi_{\mu_1}, \pi_{\mu_2}) \leq 0, \quad (\pi_t, \pi_{\mu_1}, \pi_{\mu_2}) \in J_{\mathcal{O}_N}^+ u(t, \mu_1, \mu_2).$$

A continuous function $u : \mathcal{O}_N \rightarrow \mathbb{R}$ with $u(T, \mu_1, \mu_2) = \langle \mu_1, \Phi(\cdot, \mu_2) \rangle$ is called a **viscosity supersolution** of (4.2) on \mathcal{O}_N if for every $(t, \mu_1, \mu_2) \in \mathcal{O}_N$,

$$-\pi_t - \mathcal{H}(t, \mu_1, \mu_2, \pi_{\mu_1}, \pi_{\mu_2}) \geq 0, \quad (\pi_t, \pi_{\mu_1}, \pi_{\mu_2}) \in J_{\mathcal{O}_N}^- u(t, \mu_1, \mu_2).$$

A **viscosity solution** of (4.2) is a function on \mathcal{O} that is both a subsolution and a supersolution of (4.2) on \mathcal{O}_N , for every $N \in \mathbb{N}$.

4. Master Bellman equation and viscosity solution

In what follows we show that $(t, \mu_1, \mu_2) \mapsto \mathcal{H}(t, \mu_1, \mu_2, \partial_{\mu_1} \varphi, \partial_{\mu_2} \varphi)$ is continuous on \mathcal{O}_N , for any $\varphi \in \Gamma_{\mathcal{O}_N}$.

Proposition 4.1.

Under Assumption 2.1, for every $\varphi \in \Gamma_{\mathcal{O}_N}$, the mapping

$$(t, \mu_1, \mu_2) \mapsto \mathcal{H}(t, \mu_1, \mu_2, \partial_{\mu_1} \varphi(t, \mu_1, \mu_2; \cdot), \partial_{\mu_2} \varphi(t, \mu_1, \mu_2; \cdot))$$

is continuous on \mathcal{O}_N .

Lemma 4.5.

Under the Assumptions 2.1 and 2.2, for all N , the value function ϑ is bounded on \mathcal{O}_N .

- 1 Objective of the talk
- 2 Formulation of the mean-field stochastic control problems
- 3 Dynamic programming principle
- 4 Master Bellman equation and viscosity solution
- 5 Main results**

5. Main results

Let us come to our main results: One states that **the value function ϑ is a viscosity solution of (4.2) on \mathcal{O}** , and the other shows that **the comparison theorem for (4.2) holds**.

Theorem 5.1.

Let the Assumptions 2.1 and 2.2 hold true. Then, for all $N \in \mathbb{N}$, the value function ϑ is both a viscosity sub- and a supersolution to (4.2) on \mathcal{O}_N , and so \mathcal{O} .

The proof of this theorem is rather subtle and technical, and so it is omitted here.

5. Main results

The remaining part of this section is devoted to a comparison theorem for the value function ϑ . For this purpose, we first need to introduce some definitions and notations.

Definition 5.1.

We say that a set of polynomials \mathcal{X} has the $(*)$ -property, if it satisfies

$$\text{for all } g \in \mathcal{X}, g^{(i)} \in \mathcal{X}, 0 \leq i \leq \deg(g),$$

where $g^{(i)}$ is the i -th derivative of g , and $\deg(g)$ denotes the degree of polynomial g . Let Σ be the collection of all sets of polynomials that have the $(*)$ -property.

For f a real polynomial we set

$$\mathcal{X}(f) := \bigcap_{\mathcal{X} \in \Sigma, f \in \mathcal{X}} \mathcal{X}.$$

5. Main results

We can easily check the following properties of $\mathcal{X}(f)$.

Lemma 5.1.

For every polynomial f , we have:

- (a) $\mathcal{X}(f)$ is the smallest set of polynomials with the $(*)$ -property that includes f ;
- (b) For every $g \in \mathcal{X}(f)$, $\mathcal{X}(g) \subset \mathcal{X}(f)$;
- (c) $\mathcal{X}(f)$ is finite.

Let $\Theta := \bigcup_{j=1}^{\infty} \mathcal{X}(\psi_j)$, where $\psi_j(x) = x^j$, $x \in \mathbb{R}$. Then:

- a) Θ is countable; b) $\{\psi_j\}_{j=1}^{\infty} \subset \Theta$; c) For any $f \in \Theta$, $\mathcal{X}(f) \subset \Theta$.

Let $\{f_j\}_{j=1}^{\infty}$ be an enumeration of Θ . We define the finite index set I_j :

$$I_j = \{i \mid f_i \in \mathcal{X}(f_j)\}, \quad j \geq 1.$$

Then, for all $i \in I_j$, we have $\mathcal{X}(f_i) \subset \mathcal{X}(f_j)$ and, therefore, $I_i \subset I_j$.

5. Main results

Moreover, we define for $b > 0$ and $j \in \mathbb{N}$

$$c_j(b) := \left(\sum_{k \in I_j} 2^k \right)^{-1} \left(\sum_{k \in I_j} s_k(b) \right)^{-2},$$

where $s_j(b) := 1 + \sup_{\mu \in \mathcal{M}_b} \langle \mu, f_j \rangle$. We observe that it follows from Lemma 4.3 that $1 \leq s_j(b) < \infty$, for all $j \in \mathbb{N}$. And we can see:

- Since $f_j \in \mathcal{X}(f_j)$, we have $j \in I_j$, and, thus, $c_j(b) \leq 2^{-j}$. Hence, $\sum_{j=1}^{\infty} c_j(b) \leq 1$.
- For $i \in I_j$, from $I_i \subset I_j$ we get $c_j(b) \leq c_i(b)$.
- By the definition of $s_j(b)$ and $c_j(b)$,

$$\sum_{j=1}^{\infty} c_j(b) \langle \mu, f_j \rangle^2 \leq 1, \quad \mu \in \mathcal{M}_b. \quad (5.1)$$

In the proof of comparison theorem, c_j and its properties are used to introduce a distance-like function d , in order to construct test functions.

5. Main results

To prove the comparison theorem, we need to impose an additional assumption. Let $N \geq 1$, and define

$$\mathcal{K}_N := \{ \gamma \in \mathcal{P}_2(\mathbb{R} \times U) \mid \mu := \gamma(\cdot \times U) \in \mathcal{M}_{Ne^{K^*T}} \}.$$

Note that \mathcal{K}_N is compact and $\overline{\mathcal{O}}_N \subset [0, T] \times \{ \mu = \gamma(\cdot \times U) \mid \gamma \in \mathcal{K}_N \}^2$.

Assumption 5.1(N).

There exists a constant $\kappa_0 > 0$ and a finite set $\mathcal{I} \subset \mathbb{N}$ (possibly depending on N) such that for all $s, s' \in [t, T]$, $x \in \mathbb{R}$, $u \in U$, and $\gamma, \gamma' \in \mathcal{K}_N$ satisfying $\gamma(\mathbb{R} \times \cdot) = \gamma'(\mathbb{R} \times \cdot)$,

$$|\phi(s, (x, u), \gamma) - \phi(s', (x, u), \gamma')| \leq \kappa_0(|s - s'| + \sum_{i \in \mathcal{I}} |\langle \mu - \mu', x^i \rangle|), \quad (5.2)$$

where $\mu = \gamma(\cdot \times U)$, $\mu' = \gamma'(\cdot \times U)$, for $\phi = b_1, b_2, \sigma_1, \sigma_2$, resp.

5. Main results

Remark.

Assumption 5.1(N) is a form of **Lipschitz continuity on $\overline{\mathcal{O}}_N$ w.r.t. cylindrical functions of the measure arguments**. Moreover, (5.2) also implies, that for some constant $C_{N,\mathcal{I}} > 0$,

$$|\varphi(s, (x, u), \gamma) - \varphi(s', (x, u), \gamma')| \leq K_0(|s - s'| + C_{N,\mathcal{I}}\mathcal{W}_2(\gamma, \gamma')),$$

for all $s, s' \in [t, T]$, $u \in U$, $\gamma, \gamma' \in \mathcal{K}_N$ with $\gamma(\mathbb{R} \times \cdot) = \gamma'(\mathbb{R} \times \cdot)$. This latter relation shows that, if φ is independent of (x, u) , Assumption 5.1(N) implies Assumption 2.1 on $[0, T] \times \mathcal{K}_N$.

5. Main results

Theorem 5.2. (Comparison Theorem)

We suppose that

$$b_j(t, (y, v), \gamma) = b_j(t, \gamma), \quad \sigma_j(t, (y, v), \gamma) = \sigma(t, \gamma), \quad (5.3)$$

$(t, (y, v), \gamma) \in [0, T] \times (\mathbb{R} \times U) \times \mathcal{P}_2(\mathbb{R} \times U)$, $j = 1, 2$, i.e., the coefficients b_j, σ_j are independent of (y, v) . Let Assumptions 2.2 and 5.1(N) hold on $[0, T] \times \mathcal{K}_N$. Let $u \in C(\mathcal{O}_N)$ be a viscosity subsolution to HJB equation (4.2) on \mathcal{O}_N and $v \in C(\mathcal{O}_N)$ be a viscosity supersolution to HJB equation (4.2) on \mathcal{O}_N , satisfying $u(T, \mu_1, \mu_2) \leq v(T, \mu_1, \mu_2)$, for any $(T, \mu_1, \mu_2) \in \overline{\mathcal{O}_N}$. Then $u \leq v$ on $\overline{\mathcal{O}_N}$.

Remark. Burzoni et al. [2020] consider coefficients $(b, \sigma)(t, \mu, v)$, $(t, \mu, v) \in [0, T] \times \mathcal{P}(\mathbb{R}) \times U$. While they use only deterministic control processes, we overcome this difficulty by considering our stochastic control in the law.

5. Main results

Sketch of Proof. Fix $N \in \mathbb{N}$ and let $c_j := c_j(Ne^{K^*T})$. Then, for all $(t, \mu_1, \mu_2) \in \overline{\mathcal{O}}_N$, $\mu_1, \mu_2 \in \mathcal{M}_{Ne^{K^*t}} \subset \mathcal{M}_{Ne^{K^*T}}$, it follows from (5.1) that

$$\sup_{(t, \mu_1, \mu_2) \in \overline{\mathcal{O}}_N} \sum_{j=1}^{\infty} c_j \langle \mu_i, f_j \rangle^2 \leq 1, \quad i = 1, 2.$$

We suppose that

$$\sup_{\overline{\mathcal{O}}_N} (u - v) > 0,$$

and we prove that this leads to a contradiction.

Since $u - v$ is continuous and $\overline{\mathcal{O}}_N$ is compact, the maximum

$$\ell := \max_{(t, \mu_1, \mu_2) \in \overline{\mathcal{O}}_N} ((u - v)(t, \mu_1, \mu_2) - 2\eta(T - t))$$

can be achieved and there exists sufficiently small η_0 , such that, for all $\eta \in (0, \eta_0]$, we have $\ell > 0$.

5. Main results

Now we use the standard argument of doubling variables to construct test functions for u and v .

Step 1. Doubling of variables. For $\varepsilon > 0$ and $\eta \in (0, \eta_0]$, we define

$$\begin{aligned} \phi_\varepsilon(t, \mu_1, \mu_2, s, \nu_1, \nu_2) &:= u(t, \mu_1, \mu_2) - v(s, \nu_1, \nu_2) \\ &\quad - \frac{1}{\varepsilon} \left((t - s)^2 + d(\mu_1, \nu_1) + d(\mu_2, \nu_2) \right) - \eta(T - t + T - s), \end{aligned} \quad (5.4)$$

where d is a distance-like function defined for $\mu, \nu \in \mathcal{M}_{Ne^{K^*T}}$ by the relation

$$d(\mu, \nu) := \sum_{j=1}^{\infty} c_j \langle \mu - \nu, f_j \rangle^2; \quad (5.5)$$

recall that $\{f_j\}_{j=1}^{\infty} = \Theta = \bigcup_{j=1}^{\infty} \mathcal{X}(x^j)$ (see Lemma 5.1). We observe that $d(\cdot, \cdot)$ is compatible with the weak convergence in $\mathcal{M}_{Ne^{K^*T}}$: Both generate the same topology.

5. Main results

The following corollary is a straightforward conclusion of Theorems 5.1 and 5.2.

Corollary 5.1.

Let Assumptions 2.2 and 5.1(N) hold, for all $N \geq 1$. Under the assumption that

$$b_j(t, (y, v), \gamma) = b_j(t, \gamma), \quad \sigma_j(t, (y, v), \gamma) = \sigma(t, \gamma), \quad (5.6)$$

$(t, (y, v), \gamma) \in [0, T] \times (\mathbb{R} \times U) \times \mathcal{P}_2(\mathbb{R} \times U)$, $j = 1, 2$, are independent of (y, v) , the value function ϑ is the unique viscosity solution to HJB equation (4.2) on \mathcal{O} .

Thank you very much
for your attention!