

Some aspects of the mean-field stochastic target problem

Quenched Mass Transport of Particles Towards a Target

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- 1 Introduction
- 2 Quenched mean-field SDE
- 3 Stochastic target problem
- 4 The dynamic programming PDE
- 5 Conclusion

Stochastic target problem

Consider a system described by a stochastic process $X^{t,x,\nu}$ controlled by ν and starting at x at time t .

Stochastic target problem: Look for the values x such that the system reaches a set K at a terminal time T by choosing an appropriate control ν :

Characterize the reachability sets

$$V(t) = \left\{ x \in \mathbb{R}^d : X_T^{t,x,\nu} \in K \text{ a.s. for some admissible control } \nu \right\}$$

for $t \in [0, T]$.

Motivating examples

- **Optimal reservoir management problem** (Upstream Sector in Petroleum Industry): Find the minimal amount x of liquid (e.g. water) to be injected (**fracking**) in a well at time t , to retrieve a desired amount $X_T^{t,x,\nu}$ of (shale) crude oil or gas, at time T , for some control ν (e.g. pipe dimension, pressure etc..)
- **Super-replication problem** (Finance): Find the minimal initial endowment such that there exists an investment strategy allowing the terminal wealth to be greater than a given payoff.

Other areas of application include **evacuation strategies** in crowd dynamics.

- In the Brownian diffusion case, by **the flow property**

$$X_t^{t,x} = x, \quad X_s^{t,x,\nu} = X_s^r, X_r^{t,x,\nu}, \quad \text{for any } r \in [t, s], \quad x \in \mathbb{R}^d,$$

the Geometric Dynamic Programming Principle (DPP) yields that

- $v(t, \cdot) = 1 - \mathbb{I}_{V(t)}(\cdot)$ is shown to solve an HJB equation (Soner and Touzi (2002)).
 - the function $v(t, \cdot)$ corresponding to the super-replication problem solve also an HJB equation (Soner and Touzi (2002), Bouchard et al. (2009)).
-
- In general, the minimal amount of water needed to extract shale oil/gas and the super-replication price are **too high to be afforded**.

Extension of the stochastic target problem

- Possible solution: **relax** the a.s. constraint to get a **lower** price.
- Consider injection/investment under terminal **profit & loss constraint**:

$$V_\ell(t) = \left\{ x \in \mathbb{R}^d : \mathbb{E}[\ell(X_T^{t,x,\nu})] \geq 0 \text{ for some control } \nu \right\}.$$

- Example: Take $\ell(x) = \mathbb{I}_K(x) - p$ with $p \in [0, 1]$ to get

$$V_\ell(t) = \left\{ x \in \mathbb{R}^d : \mathbb{P}(X_T^{t,x,\nu} \in K) \geq p \text{ for some control } \nu \right\}.$$

- Approach suggested in Föllmer and Leuckert (1999), then developed in Bouchard et al. (2009).
- Main idea of this last paper: use [the martingale representation theorem](#) to express the expectation constraint as an [a.s. constraint](#) on an extended process.

Our case-study

Study the stochastic target problem for **controlled nonlinear diffusion**:

$$X_s^{t,x,\nu} = \chi + \int_t^s b_u(X_u^{t,x,\nu}, \mathbb{P}_{X_u^{t,x,\nu}, \nu_u}) du + \int_t^s \sigma_u(X_u^{t,x,\nu}, \mathbb{P}_{X_u^{t,x,\nu}, \nu_u}) dB_u,$$

b, σ deterministic functions of (t, x, y, z) .

- B is a standard Brownian motion,
- χ square-integrable and \mathcal{F}_t -adapted.

Still with **the time consistent constraint** $\mathbb{E}[\ell(X_T^{t,x,\nu})] \geq 0, x \in \mathbb{R}^d$.

Extended problem: conditional law

- **Problem:** While $X^{t,\chi,\nu}$, χ square-integrable r.v., defines a flow, $X^{t,x,\nu}$, $x \in \mathbb{R}^d$ **does not have the above flow property!**

In general $X^{t,\chi,\nu} \neq X^{t,x,\nu}|_{x=\chi}$.

Consider instead a constraint of the type $\mathbb{E}[\ell(X_T^{t,\chi,\nu})] \geq 0$.

Condition on the Brownian motion B and apply the martingale representation theorem to obtain

$$\mathbb{E}[\ell(X_T^{t,\chi,\nu})] = \int \ell(x) d\mathbb{P}_{X_T^{t,\chi,\nu}}^B(x) - \int_t^T \alpha_s dB_s$$

for some control α .

The constraint $\mathbb{E}[\ell(X_T^{t,\chi,\nu})] \geq 0$ can be rewritten as

$$L(\mathbb{P}_{\tilde{\chi}_T^t, \tilde{\chi}, \tilde{\nu}}^B) \geq 0 \quad \text{or} \quad \mathbb{P}_{\tilde{\chi}_T^t, \tilde{\chi}, \tilde{\nu}}^B \in L^{-1}([0, +\infty)) \quad \text{a.s.}$$

with $\tilde{\nu} = (\nu, \alpha)$, $\tilde{\chi} = (\chi, 0)$,

$$\tilde{\chi}^{t, \tilde{\chi}, \tilde{\nu}} = (X^{t, \chi, \nu}, \int_t^\cdot \alpha_s dB_s),$$

$$L(\mu) = \int (\ell(x) - y) m(dx, dy),$$

$$m(dx, dy) := \mathbb{P}_{\tilde{\chi}_T^t, \tilde{\chi}, \tilde{\nu}}^B(dx, dy),$$

suggesting a stochastic target problem which involves $\mathbb{P}_{X_T^{t, \chi, \nu}}^B$.

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Quenched mean-field SDE

Desintegrating $\mathbb{P}_{X_T^{t,\chi,\nu}}$ w.r.t. B , the dynamics of $X^{t,\chi,\nu}$ can be written as

$$X_s^{t,\chi,\nu} = \chi + \int_t^s b_u(X_u^{t,\chi,\nu}, \mathbb{P}_{X_u^{t,\chi,\nu}}^B, \nu_u) du + \int_t^s \sigma_u(X_u^{t,\chi,\nu}, \mathbb{P}_{X_u^{t,\chi,\nu}}^B, \nu_u) dB_u.$$

Such general formulation is related to the probabilistic analysis of large scale particle systems.

In those systems, one is interested in the behavior of particles conditional on the environment ('quenched' behavior/property) (see e.g. Le Doussal and Machta (1989)).

Interpretation of the target problem

By considering a **probability law** μ as initial condition, instead of χ , our target problem can be interpreted as a **transport problem**:

What is the collection of **initial distributions** μ of a system of particles, such that the **terminal conditional law** $\mathbb{P}_{\mathcal{X}_T}^B$, given the environment (modeled by B) satisfies the **constraint**?

The reachability set reads

$$\mathcal{V}(t) = \left\{ \mu : \text{there exists } (\chi, \nu) \text{ s.t. } \mathbb{P}_{\chi}^B = \mu \text{ and } \mathbb{P}_{\mathcal{X}_T}^B \in G \text{ a.s.} \right\}.$$

Probabilistic setting

$T > 0$ fixed time horizon.

$$\Omega^\circ = \{\omega^\circ \in \mathcal{C}([0, T], \mathbb{R}^d) : \omega_0^\circ = 0\}$$

$\mathbb{F}^\circ = (\mathcal{F}_t^\circ)_{t \leq T}$ filtration generated by the canonical process
 $B(\omega^\circ) := \omega^\circ, \omega^\circ \in \Omega^\circ$.

\mathbb{P}° Wiener measure on $(\Omega^\circ, \mathcal{F}_T^\circ)$.

$\bar{\mathbb{F}}^\circ = (\bar{\mathcal{F}}_t^\circ)_{t \leq T}$ the \mathbb{P}° -completion of \mathbb{F}° .

$\Omega^! := [0, 1]^d$ endowed with σ -algebra $\mathcal{F}^! := \mathcal{B}([0, 1]^d)$ and the Lebesgue measure $\mathbb{P}^!$.

Probability space

We then define the product filtered space $(\Omega, \mathcal{F}, \mathbb{F}, \mathbb{P})$ by

- $\Omega := \Omega^\circ \times \Omega^!$,
- $\mathbb{P} = \mathbb{P}^\circ \otimes \mathbb{P}^!$,
- $\mathcal{F} = \mathcal{F}_T$ where $\mathbb{F} = (\mathcal{F}_t)_{t \leq T}$ is the completion of $(\mathcal{F}_t^\circ \otimes \mathcal{F}^!)_{t \leq T}$.

We canonically extend the random variable ξ and the process B on Ω by setting $\xi(\omega) = \xi(\omega^!)$ and $B(\omega) = B(\omega^\circ)$ for any $\omega = (\omega^\circ, \omega^!) \in \Omega$.

Advantages of this set up

Key ingredients to show the Geometric DPP and derive the HJB equation using Lions lifting argument

- If ν is \mathbb{F} -progressively measurable, then

$$\nu_s(\omega^\circ, \omega^1) = u(s, B_{\cdot \wedge s}(\omega^\circ), \xi(\omega^1)), \quad s \in [0, T],$$

with u Borel function.

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with u Borel function.

- **Jankov-von Neumann's measurable selection theorem:** there exists a **measurable** map ϑ such that (G closed set)

$$\mathbb{P}_{X_T^{\theta, \chi', \vartheta(\chi')}}^B \in G \quad \mathbb{P}^\circ - a.s. \quad \text{for } \mathfrak{P} - a.e. \quad \chi'$$

where \mathfrak{P} is the probability measure induced by $\omega^\circ \mapsto X_\theta^{t, \chi, \nu}(\omega^\circ, \cdot)$.

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Wasserstein space

We define

$$\mathcal{P}_2 := \left\{ \mu \text{ probability measure on } (\mathbb{R}^d, \mathcal{B}(\mathbb{R}^d)) \text{ s.t. } \int_{\mathbb{R}^d} |x|^2 \mu(dx) < +\infty \right\}.$$

This space is endowed with the **2-Wasserstein distance** defined by

$$\mathcal{W}_2(\mu, \mu') := \inf \left\{ \int_{\mathbb{R}^d \times \mathbb{R}^d} |x - y|^2 \pi(dy, dy) : \right. \\ \left. \text{s.t. } \pi(\cdot \times \mathbb{R}^d) = \mu \text{ and } \pi(\mathbb{R}^d \times \cdot) = \mu' \right\}^{\frac{1}{2}},$$

for $\mu, \mu' \in \mathcal{P}_2$. For later use, we also define the collection $\mathcal{P}_2^{\bar{\mathbb{F}}^0}$ of $\bar{\mathbb{F}}^0$ -adapted continuous \mathcal{P}_2 -valued processes.

Controlled quenched diffusion

Let U be a closed subset of \mathbb{R}^q for some $q \geq 1$ and \mathcal{U} the set of U -valued \mathbb{F} -progressive processes.

Given

- $\theta \in \bar{\mathcal{T}}^\circ$ (the set of $[0, T]$ -valued $\bar{\mathbb{F}}^\circ$ -stopping times),
- $\chi \in L^2(\Omega, \mathcal{F}_\theta, \mathbb{P}; \mathbb{R}^d)$,
- $\nu \in \mathcal{U}$,

we let $X^{\theta, \chi, \nu}$ denote the solution of

$$X = \mathbb{E}[\chi | \mathcal{F}_{\theta \wedge \cdot}] + \int_{\theta}^{\theta \vee \cdot} b_s(X_s, \mathbb{P}_{X_s}^B, \nu_s) ds + \int_{\theta}^{\theta \vee \cdot} a_s(X_s, \mathbb{P}_{X_s}^B, \nu_s) dB_s,$$

Existence, uniqueness and stability

We suppose that b, a are continuous, bounded and there exists a constant L such that

$$|b_t(x, \mu, \cdot) - b_t(x', \mu', \cdot)| + |a_t(x, \mu, \cdot) - a_t(x', \mu', \cdot)| \leq L(|x - x'| + \mathcal{W}_2(\mu, \mu'))$$

for all $t \in [0, T]$, $x, x' \in \mathbb{R}^d$ and $\mu, \mu' \in \mathcal{P}_2$.

Proposition

For all $\theta \in \bar{T}^\circ$, $\nu \in \mathcal{U}$ and $\chi \in L^2(\mathcal{F}_\theta)$, the SDE admits a unique strong solution $X^{\theta, \chi, \nu}$, and it satisfies

$$\mathbb{E} \left[\sup_{[0, T]} |X^{\theta, \chi, \nu}|^2 \right] < +\infty,$$

$$\mathbb{P}_{X_T^t, \chi, \nu}^B = \mathbb{P}_{X_T^\theta, X_\theta^t, \chi, \nu, \nu}^B \quad (\text{Flow property}).$$

Moreover, if $(t_n, \chi_n) \rightarrow (t, \chi)$ and $(\nu^n)_n \subset \mathcal{U}$ converges to ν $dt \otimes d\mathbb{P}$ -a.e., then

$$\lim_{n \rightarrow \infty} \mathbb{E}[\mathcal{W}_2(\mathbb{P}_{X_T^{t_n}, \chi_n, \nu^n}^B, \mathbb{P}_{X_T^t, \chi, \nu}^B)^2] = 0.$$

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First formulation

Look for **the set of initial measures** for the conditional law \mathbb{P}_χ^B such that the **terminal conditional law** of $X_T^{t,\chi,\nu}$ given B **belongs to a fixed closed subset** G of \mathcal{P}_2 :

$$\mathcal{V}(t) = \left\{ \mu \in \mathcal{P}_2 : \text{there exists } (\chi, \nu) \in L^2(\mathcal{F}_t) \times \mathcal{U} \text{ s.t. } \mathbb{P}_\chi^B = \mu \text{ and } \mathbb{P}_{X_T^{t,\chi,\nu}}^B \in G \right\}.$$

This formulation is **not convenient** for setting a Geometric DPP:

- In $\mathcal{V}(t)$ only the probability distribution μ **should matter** and not a particular representation χ .

Strong formulation

The following strong formulation allows to **take any representing random variable χ** for μ .

Proposition

A measure $\mu \in \mathcal{P}_2$ belongs to $\mathcal{V}(t)$ if and only if for all $\chi \in L^2(\mathcal{F}_t)$ such that $\mathbb{P}_\chi^B = \mu$ there exists $\nu \in \mathcal{U}$ for which $\mathbb{P}_{\chi_T^t, \chi, \nu}^B \in G$:

$$\mathcal{V}(t) = \left\{ \mu \in \mathcal{P}_2 : \forall \chi \in L^2(\mathcal{F}_t) \text{ s.t. } \mathbb{P}_\chi^B = \mu \exists \nu \in \mathcal{U} \text{ for which } \mathbb{P}_{\chi_T^t, \chi, \nu}^B \in G \right\}.$$

This defines a mass transport problem towards a given target along the path of a mean-field diffusion.

Dynamic programming principle

Theorem

Fix $t \in [0, T]$ and $\theta \in \bar{\mathcal{T}}^\circ$ with values in $[t, T]$. Then,

$$\mathcal{V}(t) = \left\{ \mu \in \mathcal{P}_2 : \exists (\chi, \nu) \in L^2(\mathcal{F}_t) \times \mathcal{U} \text{ s.t. } \mathbb{P}_\chi^B = \mu \text{ and } \mathbb{P}_{X_\theta^{t, \chi, \nu}}^B \in \mathcal{V}(\theta) \right\}.$$

Note that this DPP holds only for **stopping times in $\bar{\mathcal{T}}^\circ$** i.e. stopping time w.r.t. the Brownian filtration.

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The value function

Let $v : [0, T] \times \mathcal{P}_2 \rightarrow \mathbb{R}$ be the indicator function of the complement of the reachability set \mathcal{V} :

$$v(t, \mu) = 1 - \mathbb{I}_{\mathcal{V}(t)}(\mu), \quad (t, \mu) \in [0, T] \times \mathcal{P}_2.$$

Aim: provide a **characterization** of v as a (discontinuous viscosity) solution of a **fully non-linear second order parabolic partial differential equation**.

Lift on \mathcal{P}_2

Aim: define derivatives for functions defined on \mathcal{P}_2 .

- **Issue:** \mathcal{P}_2 is not a vector space.

Possible approach: [Lions Lifting](#)

For a function $w : \mathcal{P}_2 \rightarrow \mathbb{R}$, we define its **lift** as $W : L^2(\Omega^1, \mathcal{F}^1, \mathbb{P}^1; \mathbb{R}^d) \rightarrow \mathbb{R}$ such that

$$W(X) = w(\mathbb{P}_X), \quad \text{for all } X \in L^2(\Omega^1, \mathcal{F}^1, \mathbb{P}^1; \mathbb{R}^d).$$

Allows to consider functions defined on the [Hilbert space](#) $L^2(\Omega^1, \mathcal{F}^1, \mathbb{P}^1; \mathbb{R}^d)$.

Derivatives on \mathcal{P}_2

We then say that w is **Fréchet differentiable** (resp. \mathcal{C}^1) on \mathcal{P}_2 if its lift W is (resp. continuously) **Fréchet differentiable** on $L_2(\Omega^!, \mathcal{F}^!, \mathbb{P}^!; \mathbb{R}^d)$.

Then $DW(X) \in L^2(\Omega^!, \mathcal{F}^!, \mathbb{P}^!; \mathbb{R}^d)$ admits the **representation**

$$DW(X) = \partial_\mu w(\mathbb{P}_X)(X)$$

with $\partial_\mu w(\mathbb{P}_X) : \mathbb{R}^d \rightarrow \mathbb{R}^d$ measurable map, called the derivative of w at \mathbb{P}_X .

We have $\partial_\mu w(\mu) \in L^2(\mathbb{R}^d, \mathcal{B}(\mathbb{R}^d), \mu; \mathbb{R}^d)$ for $\mu \in \mathcal{P}_2$.

We denote by $\partial_x \partial_\mu w(\mu)(x)$ the gradient of $x \in \mathbb{R}^d \mapsto \partial_\mu w(\mu)(x)$.

We have the following **identification**

$$\mathbb{E} \left[D^2 W(X)(YZ)YZ^\top \right] = \mathbb{E} \left[\text{Tr} \left(\partial_x \partial_\mu w(\mu)(X)YY^\top \right) \right] \quad (1)$$

for any $Y \in L^2(\Omega^!, \mathcal{F}^!, \mathbb{P}^!; \mathbb{R}^{d \times d})$, $Z \sim N(0, I_d)$ and $Z \perp\!\!\!\perp (X, Y)$

Chain rule

Proposition

Let $w \in \mathcal{C}_b^{1,2}([0, T] \times \mathcal{P}_2)$. Given $(t, \chi, \nu) \in [0, T] \times L^2(\mathcal{F}_t) \times \mathcal{U}$, set $X = X^{t, \chi, \nu}$. Then,

$$\begin{aligned} w(s, \mathbb{P}_{X_s}^B) &= w(t, \mathbb{P}_{X_t}^B) \\ &+ \int_t^s \mathbb{E}_B \left[\partial_t w(r, \mathbb{P}_{X_r}^B) + \partial_\mu w(r, \mathbb{P}_{X_r}^B)(X_r) b_r(X_r, \mathbb{P}_{X_r}^B, \nu_r) \right] dr \\ &+ \frac{1}{2} \int_t^s \mathbb{E}_B \left[\text{Tr} \left(\partial_x \partial_\mu w(r, \mathbb{P}_{X_r}^B)(X_r) a_r a_r^\top(X_r, \mathbb{P}_{X_r}^B, \nu_r) \right) \right] dr \\ &+ \int_t^s \mathbb{E}_B \left[\partial_\mu w(r, \mathbb{P}_{X_r}^B)(X_r) a_r(X_r, \mathbb{P}_{X_r}^B, \nu_r) \right] dB_r \end{aligned}$$

for all $s \in [t, T]$. Here $\mathbb{E}_B[\cdot]$ means conditioning w.r.t. $(B_r, r \leq T)$.

Chain rule on L^2

Given $X \in L^2(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)$, denote $W(t, X)$ the r.v.
 $\omega^0 \in \Omega^0 \mapsto W(t, X(\omega^0, \cdot))$ a r.v. in $L^2(\Omega^0, \mathcal{F}^0, \mathbb{P}^0; \mathbb{R}^d)$.

Corollary

Let $W : [0, T] \times L^2(\Omega^0, \mathcal{F}^0, \mathbb{P}^0; \mathbb{R}^d) \rightarrow \mathbb{R}$ be $C_b^{1,2}$. Set $X = X^{t, \chi, \nu}$. Then,

$$\begin{aligned} W(s, X_s) &= W(t, \chi) \\ &+ \int_t^s \mathbb{E}_B \left[\partial_t W(r, X_r) + DW(r, X_r) b_r(X_r, \mathbb{P}_{X_r}^B, \nu_r) \right] dr \\ &+ \frac{1}{2} \int_t^s \mathbb{E}_B \left[D^2 W(r, X_r)(X_r)(a_r Z)(a_r Z)^\top(X_r, \mathbb{P}_{X_r}^B, \nu_r) \right] dr \\ &+ \int_t^s \mathbb{E}_B \left[DW(r, X_r) a_r(X_r, \mathbb{P}_{X_r}^B, \nu_r) \right] dB_r \end{aligned}$$

for all $s \in [0, T]$, with $Z \sim N(0, I_d)$ and $Z \perp \chi, B$.

A 'quenched' PDE

We show that $V : [0, T] \times L^2(\Omega', \mathcal{F}', \mathbb{P}'; \mathbb{R}^d) \rightarrow \mathbb{R}$ (lift of v) is a **viscosity solution** on $[0, T] \times L^2(\Omega', \mathcal{F}', \mathbb{P}'; \mathbb{R}^d)$ of the **quenched PDE**

$$-\partial_t W(t, \xi) + \mathcal{H}(t, \xi, DW(t, \xi), D^2 W(t, \xi)) = 0.$$

where $\mathcal{H} = \lim_{\varepsilon \rightarrow 0^+} \mathcal{H}_\varepsilon$ with

$$\mathcal{L}_t^u(\xi, P, Q) := \mathbb{E}_B \left[b_t^\top(\xi, \mathbb{P}_\xi, u)P + \frac{1}{2}Q(a_t(\xi, \mathbb{P}_\xi, u)Z)a_t(\xi, \mathbb{P}_\xi, u)Z \right]$$

$$\mathcal{H}_\varepsilon(t, \xi, P, Q) := \sup_{u \in \mathcal{N}_\varepsilon(t, \xi, P)} \left\{ -\mathcal{L}_t^u(\xi, P, Q) \right\}$$

$$\mathcal{N}_\varepsilon(t, \xi, P) := \left\{ u \in L^0(\Omega, \mathcal{F}, \mathbb{P}; U) : |\mathbb{E}_B[a_t(\chi, \mathbb{P}_\xi, u)P]| \leq \varepsilon \right\},$$

where $\mathbb{E}_B[\cdot]$ means conditioning w.r.t. $(B_r, r \leq T)$.

Continuity assumption

We need the following **assumption**. It ensures the existence of a **regular feedback control** 'close' to the **kernel** \mathcal{N}_0 .

Continuity Assumption: Let \mathcal{O} be an open subset of $[0, T] \times [L^2(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)]^2$ such that $\mathcal{N}_0 \neq \emptyset$ on \mathcal{O} . Then, for every $\varepsilon > 0$, $(t_0, \chi_0, P_0) \in \mathcal{O}$ and $u_0 \in \mathcal{N}_0(t_0, \chi_0, P_0)$, there exists **an open neighborhood** \mathcal{O}' of (t_0, χ_0, P_0) and **a measurable map** $\hat{u} : [0, T] \times \mathbb{R}^d \times \mathbb{R}^d \times \Omega' \rightarrow U$ such that:

- (i) $\mathbb{E}_B[|\hat{u}_{t_0}(\chi_0, P_0, \xi) - u_0|] \leq \varepsilon$,
- (ii) there exists $C > 0$ for which

$$\mathbb{E}[|\hat{u}_t(\chi, P, \xi) - \hat{u}_t(\chi', P', \xi)|^2] \leq C\mathbb{E}[|\chi - \chi'|^2 + W_2^2(\mathbb{P}_P, \mathbb{P}_{P'})]$$

for all $(t, \chi, P), (t, \chi', P') \in \mathcal{O}'$,

(iii) $\hat{u}_t(\chi, P, \xi) \in \mathcal{N}_0(t, \chi, P) \mathbb{P}^\circ - \text{a.e.}$, for all $(t, \chi, P) \in \mathcal{O}'$,

Viscosity property

We also suppose that there exists a constant C and a function $m : \mathbb{R}_+ \rightarrow \mathbb{R}$ such that $m(t) \xrightarrow[t \rightarrow 0]{} 0$ and

$$|b_t(x, \mu, u) - b_{t'}(x, \mu, u')| + |a_t(x, \mu, u) - a_{t'}(x, \mu, u')| \leq m(t - t') + C|u - u'|.$$

for all $t, t' \in [0, T]$, $x \in \mathbb{R}^d$, $\mu \in \mathcal{P}_2$ and $u, u' \in U$.

Theorem

The function V is a *viscosity supersolution* of the HJB equation. If in addition the **Continuity Assumption** holds, then V is also a *viscosity subsolution* of the HJB equation.

Parabolic boundary conditions

Define the function g by

$$g(\xi) = 1 - \mathbb{I}_G(\mathbb{P}_\xi), \quad \xi \in L^2(\Omega^1, \mathcal{F}^1, \mathbb{P}^1; \mathbb{R}^d)$$

and g_* and g^* its **lower** and **upper** semi-continuous envelopes.

Theorem

*Under **(H1)**, the function V satisfies*

$$V^*(T, \cdot) = g^* \quad \text{and} \quad V_*(T, \cdot) = g_*$$

on $L^2(\Omega^1, \mathcal{F}^1, \mathbb{P}^1; \mathbb{R}^d)$.

Outline

- 1 Introduction
- 2 Quenched mean-field SDE
- 3 Stochastic target problem
- 4 The dynamic programming PDE
- 5 Conclusion**





Conclusion and perspectives

- Introduced a **seemingly new stochastic target problem** with potential **financial** and **engineering** applications.
- Obtained a **dynamic programming principle**
- Obtained a **random PDE** and derived some of its properties





Extensions and open problems

- Uniqueness (or a comparison result) for the PDE
- Target problem for \mathbb{P}_{X_T} (unconditional law)
- Numerics for the quenched PDE.
- Processes quenched by other environments such as jump processes, long memory processes (fractional BM etc..)

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Thank You!