

# Pontryagin maximum principle for the deterministic mean field type optimal control problem via the Lagrangian approach

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## Mean field type control problem. Dynamics

Let

- ▶  $\mathbb{R}^d$  be a phase space for each agent;
- ▶  $f(t, x, m, u)$ , where  $t \in [0, T]$ ,  $x \in \mathbb{R}^d$ ,  $m$  is a probability on  $\mathbb{R}^d$ ,  $u \in U$  is a control, be a nonlocal **velocity field**.

Dynamics of distribution of agents satisfies in the distributional sense the **nonlocal continuity equation**:

$$\frac{\partial}{\partial t} m(t) + \operatorname{div}(f(t, x, m(t), u(t, x))m(t)) = 0.$$

In particular, the dynamics of each agent obeys the ODE:

$$\dot{x} = f(t, x, m(t), u(t, x)).$$

## Mean field type control problem.

- ▶ The agents play **cooperatively** to minimize the **averaged** payoff.
- ▶ The payoff of each agent is equal to

$$\sigma(x(T), m(T)) + \int_0^T f_0(t, x(t), m(t), u(t, x(t))) dt.$$

## Notation

- ▶ If  $(X, \rho_X)$  is a Polish space, then  $\mathcal{B}(X)$  denotes the Borel  $\sigma$ -algebra on  $X$ .
- ▶  $\mathcal{P}(X)$  is the set of Borel probabilities on  $X$ .

## Push-forward measure

Assume that

- ▶  $(\Omega, \mathcal{F}), (\Omega', \mathcal{F}')$  are measurable spaces,
- ▶  $\mathbb{P}$  is a probability on  $\mathcal{F}$ ,
- ▶  $\xi : \Omega \rightarrow \Omega'$  is measurable function.

A probability  $\xi\#\mathbb{P}$  on  $\mathcal{F}'$  defined by the rule: for  $E \in \mathcal{F}'$ ,

$$(\xi\#\mathbb{P})(E) \triangleq \mathbb{P}(\xi^{-1}(E))$$

is called a **push-forward measure**.

## Notation. Space of probabilities

- ▶ If  $(X, \rho_X)$  is a Polish space,  $p \geq 1$ , then  $\mathcal{P}^p(X)$  is the set of probabilities on  $X$  with the finite  $p$ -th moment, i.e.,  $m \in \mathcal{P}^p(X)$  iff, for some (equivalently, any)  $x_* \in X$ ,

$$\mathcal{M}_p^p(m) \triangleq \int_X \rho_X^p(x, x_*) m(dx) < \infty.$$

- ▶ Distance on  $\mathcal{P}^p(X)$ : if  $m_1, m_2 \in \mathcal{P}^p(X)$ , then

$$W_p(m_1, m_2) \triangleq \inf \left[ \int_{X \times X} \rho_X^p(x_1, x_2) \pi(dx_1 dx_2) : \pi \in \Pi(m_1, m_2) \right]^{1/p},$$

where  $\Pi(m_1, m_2)$  is the set of probabilities  $\pi$  on  $X \times X$  such that, for any measurable  $E \subset X$ ,  $\pi(E \times X) = m_1(E)$ ,  $\pi(X \times E) = m_2(E)$ .

## Notation. State and controls

- ▶ Space of curves:  $\Gamma = C([0, T]; \mathbb{R}^d)$ .
- ▶ Space of curves in the costate space:  $\Gamma^* \triangleq C([0, T]; \mathbb{R}^{d,*})$ .
- ▶ Evaluation operator: for  $\gamma \in \Gamma$ ,

$$e_t(\gamma) = \gamma(t).$$

- ▶ Space of controls:  $\mathcal{U}^p \triangleq L^p([0, T], \mathcal{B}([0, T]), \lambda; U)$ , where  $\lambda$  stands for the Lebesgue measure.

## Lagrangian approach

- ▶  $(\Omega, \mathcal{F}, \mathbb{P})$  is a standard probability space.
- ▶ **Control process:**  $(X, u_L)$ , where  $X \in L^p(\Omega, \mathcal{F}, \mathbb{P}; \Gamma)$ ,  $u_L \in L^p(\Omega, \mathcal{F}, \mathbb{P}; \mathcal{U}^p)$ .
- ▶ **Dynamics:**

$$\frac{d}{dt}X(t, \omega) = f(t, X(t, \omega), X(t)\# \mathbb{P}, u_L(t, \omega)).$$

- ▶ **Initial condition:**  $X(0)\# \mathbb{P} = m_0$ .
- ▶ **Payoff:**

$$J_L(X, u_L) \triangleq \int_{\Omega} \sigma(X(T, \omega), X(T)\# \mathbb{P}) \mathbb{P}(d\omega) + \int_{\Omega} \int_0^T f_0(t, X(t, \omega), X(t)\# \mathbb{P}, u_L(t, \omega)) dt \mathbb{P}(d\omega).$$



## Kantorovich approach

- ▶ **Control process:**  $(\eta, u_K)$ , where  $\eta \in \mathcal{P}^p(\Gamma)$ ,  $u_K \in L^p(\Gamma, \mathcal{B}(\Gamma), \eta; \mathcal{U}^p)$ .
- ▶ **Feasibility:** for  $\eta$ -a.e.  $\gamma \in \Gamma$ ,

$$\frac{d}{dt}\gamma(t) = f(t, \gamma(t), e_t \# \eta, u_K(t, \gamma)),$$

where  $e_t(\gamma) = \gamma(t)$ ,  $(e_t \# \eta)(E) = \eta\{\gamma \in \Gamma : \gamma(t) \in E\}$ .

- ▶ **Initial condition:**  $e_0 \# \eta = m_0$ .
- ▶ **Payoff:**

$$J_K(\eta, u_K) \triangleq \int_{\Gamma} \sigma(\gamma(T), e_T \# \eta) \eta(d\gamma) + \int_{\Gamma} \int_0^T f_0(t, \gamma(t), e_t \# \eta, u_K(t, \gamma)) dt \eta(d\gamma).$$

## Eulerian approach

- ▶ **Control process:**  $(m(\cdot), u_E)$ , where  $m(t)$  is a probability on  $\mathbb{R}^d$ ,  $u_E : [0, T] \times \mathbb{R}^d \rightarrow U$ .
- ▶ **Dynamics:**  $m(\cdot)$  is a distributional solution of the nonlocal continuity equation:

$$\partial_t m(t) + \operatorname{div}(v_E(t, x)m(t)) = 0,$$

for  $v_E(t, x) = f(t, x, m(t), u_E(t, x))$ .

- ▶ **Initial condition:**  $m(0) = m_0$ .
- ▶ **Payoff:**

$$\begin{aligned} J_E(\mu, u_E) &\triangleq \int_{\mathbb{R}^d} \sigma(x, m(T))m(T, dx) \\ &+ \int_0^T \int_{\mathbb{R}^d} f_0(t, x, m(t), u_E(t, x))m(t, dx)dt. \end{aligned}$$

## Continuity equation

$m(\cdot)$  is a distributional solution of the nonlocal continuity equation:

$$\partial_t m(t) + \operatorname{div}(v(t, x)m(t)) = 0,$$

iff, for every  $\varphi \in C_c^\infty((0, T) \times \mathbb{R}^d)$ ,

$$\int_0^T \int_{\mathbb{R}^d} [\partial_t \varphi(t, x) + \nabla_x \varphi(t, x)v(t, x)] m(t, x) dt = 0.$$

# Intrinsic derivative

## Definition

Let  $\Phi : \mathcal{P}^p(\mathbb{R}^d) \rightarrow \mathbb{R}$ . A function  $\frac{\delta\Phi}{\delta m} : \mathcal{P}^p(\mathbb{R}^d) \times \mathbb{R}^d \rightarrow \mathbb{R}$  is a **flat derivative** iff, for any  $m' \in \mathcal{P}^p(\mathbb{R}^d)$ ,

$$\begin{aligned} \lim_{s \downarrow 0} \frac{\Phi((1-s)m + sm') - \Phi(m)}{s} \\ = \int_{\mathbb{R}^d} \frac{\delta\Phi}{\delta m}(m, y) [m'(dy) - m(dy)]. \end{aligned}$$

# Intrinsic derivative

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## Definition

The function  $\nabla_m \Phi$  defined by the rule

$$\nabla_m \Phi(m, y) \triangleq \nabla_y \frac{\delta\Phi}{\delta m}(m, y)$$

is called an **intrinsic derivative** of the function  $\Phi$ .

## Assumptions

- ▶  $p > 1$ ;
- ▶  $U$  is a closed subset of a separable Banach space;
- ▶ the functions  $f$ ,  $f_0$  and  $\sigma$  are continuous;
- ▶ there exists a constant  $C_\infty$  such that

$$\|f(t, x, m, u)\| \leq C_\infty(1 + \|x\| + \mathcal{M}_p(m) + \|u\|),$$

$$|f_0(t, x, m, u)| \leq C_\infty(1 + \|x\|^p + \mathcal{M}_p^p(m) + \|u\|^p),$$

$$|\sigma(x, m)| \leq C_\infty(1 + \|x\|^p + \mathcal{M}_p^p(m));$$

- ▶ the functions  $f$  and  $f_0$  are continuous w.r.t. time variable with the same modulus of continuity  $\varsigma(\cdot)$ ;
- ▶ the function  $f$  is continuously differentiable w.r.t.  $x$  and  $m$ ; its derivatives  $\nabla_x f$  and  $\nabla_m f$  are bounded by constants  $C_x$  and  $C_m$  respectively;

## Assumptions

- ▶ the function  $f_0$  is continuously differentiable w.r.t.  $x$  and  $m$ ; the derivatives  $\nabla_x f_0$  and  $\nabla_m f_0$  satisfy the following growth conditions with constants  $C_x^0, C_m^0$ :

$$\|\nabla_x f_0(t, x, m, u)\|^q \leq C_x^0(1 + \|x\|^p + \mathcal{M}_p^p(m) + \|u\|^p),$$

$$\|\nabla_m f_0(t, x, m, y, u)\|^q \leq C_m^0(1 + \|x\|^p + \|y\|^p + \mathcal{M}_p^p(m) + \|u\|^p);$$

- ▶ the terminal payoff  $\sigma$  is continuously differentiable; the functions  $\nabla_x \sigma$  and  $\nabla_m \sigma$  satisfy the following estimates with some nonnegative constants  $C_x^\sigma, C_m^\sigma$ :

$$\|\nabla_x \sigma(x, m)\|^q \leq C_x^\sigma(1 + \|x\|^p + \mathcal{M}_p^p(m)),$$

$$\|\nabla_m \sigma(x, m, y)\|^q \leq C_m^\sigma(1 + \|x\|^p + \|y\|^p + \mathcal{M}_p^p(m)).$$

## Lagrangian strong local $L^p$ -minimizer

A Lagrangian control process  $(X^*, u_L^*)$  is a **strong local  $L^p$ -minimizer** at  $m_0$  if there exists  $\varepsilon > 0$  satisfying the following condition: for any admissible  $(X, u_L)$  such that  $\|X - X^*\|_{L^p} \leq \varepsilon$ ,

$$J_L(X^*, u_L^*) \leq J_L(X, u_L).$$



## Lagrangian strong local $W_p$ -minimizer

A Lagrangian control process  $(X^*, u_L^*)$  is a **strong local  $W_p$ -minimizer** at  $m_0$  if there exists  $\varepsilon > 0$  satisfying the following condition: for every admissible  $(X, u_L)$  such that  $W_p(X \# \mathbb{P}, X^* \# \mathbb{P}) \leq \varepsilon$ ,

$$J_L(X^*, u_L^*) \leq J_L(X, u_L).$$

## Pontryagin local $L^p$ -minimizer for the Lagrangian approach

A Lagrangian control process  $(X^*, u_L^*)$  is a **Pontryagin local  $L^p$ -minimizer** at  $m_0$  if there exists  $\varepsilon > 0$  such that inequality

$$J_L(X^*, u_L^*) \leq J_L(X, u_L).$$

holds true for each admissible  $(X, u_L)$  satisfying

- ▶  $\|X - X^*\|_{L^p} \leq \varepsilon,$
- ▶  $(\lambda \otimes \mathbb{P})\{(t, \omega) \in [0, T] \times \Omega : u^*(t, \omega) \neq u(t, \omega)\} \leq \varepsilon.$

## Pontryagin local $W_p$ -minimizer for the Lagrangian approach

A Lagrangian control process  $(X^*, u_L^*)$  is a **Pontryagin local  $W_p$ -minimizer** at  $m_0$  if there exists  $\varepsilon > 0$  such that

$$J_L(X^*, u_L^*) \leq J_L(X, u_L).$$

holds true for each admissible  $(X, u_L)$  satisfying

- ▶  $W_p(X \# \mathbb{P}, X^* \# \mathbb{P}) \leq \varepsilon,$
- ▶  $(\lambda \otimes \mathbb{P})\{(t, \omega) \in [0, T] \times \Omega : u^*(t, \omega) \neq u(t, \omega)\} \leq \varepsilon.$

## Local minima within Lagrangian approach

- ▶ Every strong minimizer is a Pontryagin one.
- ▶ Every  $W_p$  minimizer is a  $L^p$ -minimizer.

The Pontryagin  $L^p$ -minimizer is the weakest minimizer!

Pontryagin maximum principle for the Lagrangian approach

## Regular points

A point  $t \in [0, T]$  is called **regular** for the process  $(X, u)$  if

▶  $\|u(t)\|_{L^p} < \infty,$



$$\lim_{h \downarrow 0} \mathbb{E} \left\| \frac{1}{h} \int_t^{t+h} f(\tau, X(\tau), X(\tau) \# \mathbb{P}, u(\tau)) d\tau - f(t, X(t), X(t) \# \mathbb{P}, u(t)) \right\|^p = 0,$$



$$\lim_{h \downarrow 0} \mathbb{E} \left| \frac{1}{h} \int_t^{t+h} f_0(\tau, X(\tau), X(\tau) \# \mathbb{P}, u(\tau)) d\tau - f_0(t, X(t), X(t) \# \mathbb{P}, u(t)) \right| = 0.$$

## Regular points

Almost every  $t \in [0, T]$  are regular.

# Pontryagin function for the Lagrangian PMP

- ▶ **local Pontryagin function:** for  $t \in [0, T]$ ,  $x \in \mathbb{R}^d$ ,  
 $m \in \mathcal{P}^p(\mathbb{R}^d)$ ,  $\psi \in \mathbb{R}^{d,*}$ ,  $u \in U$ ,

$$H(t, x, m, \psi, u) \triangleq \psi f(t, x, m, u) - f_0(t, x, m, u);$$

- ▶ **global Pontryagin function:** for  $t \in [0, T]$ ,  
 $X \in L^p(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)$ ,  $\Psi \in L^q(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^{d,*})$ ,  
 $u \in L^p(\Omega, \mathcal{F}, \mathbb{P}; U)$ ,

$$\mathbb{H}(t, X, \Psi, u) \triangleq \mathbb{E}H(t, X, X \# \mathbb{P}, \Psi, u).$$



## Pontryagin maximum principle in the Lagrangian form

Let  $(\Omega, \mathcal{F}, \mathbb{P})$  be a standard probability space, a Lagrangian process  $(X^*, u^*)$  be a Pontryagin local  $L^p$ -minimizer. Then there exists a function  $\Psi \in L^q(\Omega, \mathcal{F}, \mathbb{P}; \Gamma^*)$  such that

- ▶ the costate equation,
- ▶ the transversality condition,
- ▶ the maximization conditions

hold true.

## Costate equation

For  $\mathbb{P}$ -a.e.  $\omega \in \Omega$ ,  $\Psi(\cdot, \omega)$  solves

$$\begin{aligned} \frac{d}{dt} \Psi(t, \omega) &= -\Psi(t, \omega) \nabla_x f(t, X^*(t, \omega), X^*(t) \# \mathbb{P}, u^*(t, \omega)) \\ &\quad + \nabla_x f_0(t, X^*(t, \omega), X^*(t) \# \mathbb{P}, u^*(t)) \\ &\quad - \int_{\Omega} \Psi(t, \omega') \nabla_m f(t, X^*(t, \omega'), X^*(t) \# \mathbb{P}, \\ &\quad \quad \quad X^*(t, \omega), u^*(t, \omega')) \mathbb{P}(d\omega') \\ &\quad + \int_{\Omega} \nabla_m f_0(t, X^*(t, \omega'), X^*(t) \# \mathbb{P}, \\ &\quad \quad \quad X^*(t, \omega), u^*(t, \omega')) \mathbb{P}(d\omega'). \end{aligned}$$

## Transversality condition

$$\begin{aligned}\Psi(T, \omega) &= -\nabla_x \sigma(X^*(T, \omega), X^*(T) \# \mathbb{P}) \\ &\quad - \int_{\Omega} \nabla_m \sigma(X^*(T, \omega'), X^*(T) \# \mathbb{P}, X^*(T, \omega)) \mathbb{P}(d\omega')\end{aligned}$$

$\mathbb{P}$ -a.s.

## Maximization of the Hamiltonian condition

- ▶ **global form:** for each regular point  $t \in [0, T]$ ,

$$\mathbb{H}(s, X^*(s), \Psi(s), u^*(s)) = \max_{\nu \in L^p(\Omega, \mathcal{F}, \mathbb{P}; U)} \mathbb{H}(s, X^*(s), \Psi(s), \nu)$$

- ▶ **local form:** for each regular point  $t \in [0, T]$ ,

$$\begin{aligned} H(s, X^*(s), X^*(s) \# \mathbb{P}, \Psi(s), u^*(s)) \\ = \max_{u \in U} H(s, X^*(s), X^*(s) \# \mathbb{P}, \Psi(s), u) \quad \mathbb{P}\text{-a.s.} \end{aligned}$$

## Hamiltonian form of the costate equation

$$\begin{aligned}\frac{d}{dt}X^*(t) &= \nabla_{\Psi}H(t, X^*(t), \Psi(t), u^*(t)), \\ \frac{d}{dt}\Psi(t) &= -\nabla_X H(t, X^*(t), \Psi(t), u^*(t)), \\ X(0) &= X_0, \\ \Psi(T) &= -\nabla_X \Sigma(X^*(T)).\end{aligned}$$

Here,

- ▶  $\nabla_X H$ ,  $\nabla_{\Psi} H$  stands for the derivatives w.r.t.  $X \in L^p(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)$  and  $\Psi \in L^q(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)$ ,
- ▶  $\Sigma(X) \triangleq \mathbb{E}\sigma(X, X\#\mathbb{P})$ .

## Method of proof: spike variation

Let  $\nu \in L^p(\Omega, \mathcal{F}, \mathbb{P}; U)$  and let  $s \in [0, T]$  be a regular point.

Perturbated control:

$$u_\nu^h(t, \omega) \triangleq \begin{cases} u^*(t, \omega), & t \in [0, s), \\ \nu(\omega), & t \in [s, s+h), \\ u^*(t, \omega) & t \in [s+h, T]. \end{cases}$$

Perturbated motion:  $Z_\nu^h \in L^p(\Omega, \mathcal{F}, \mathbb{P}; \Gamma)$  satisfies

$$\frac{d}{dt} Z_\nu^h(t, \omega) = f(t, Z_\nu^h(t, \omega), Z_\nu^h(t) \# \mathbb{P}, u_\nu^h(t, \omega)),$$

$$Z_\nu^h(0, \omega) = X_0(\omega).$$

## Sequence of perturbed controls

There exists a sequence  $\{h_n\}_{n=1}^\infty$  such that

1.  $\{Z_\nu^{h_n}(t, \omega)\}_{n=1}^\infty$  converges to  $X^*(t, \omega)$  for  $\lambda \otimes \mathbb{P}$ -a.e.  $(t, \omega) \in [s, T] \times \Omega$ ;
2.  $\{Z_\nu^{h_n}(T, \omega)\}_{n=1}^\infty$  converges to  $X^*(T, \omega)$  for  $\mathbb{P}$ -a.e.  $\omega \in \Omega$ .

## Variational equation. Notation

- ▶ derivative w.r.t.  $x$ :

$$f_x^*(t, \omega) \triangleq \nabla_x f(t, X^*(t, \omega), X^*(t) \# \mathbb{P}, u^*(t, \omega));$$

- ▶ derivative w.r.t.  $m$ :

$$f_m^*(t, \omega, \omega') \triangleq \nabla_m f(t, X^*(t, \omega), X^*(t) \# \mathbb{P}, X^*(t, \omega'), u^*(t, \omega));$$

- ▶ inner product:

$$\langle f_m^*, Y_\nu \rangle(t, \omega) \triangleq \int_{\Omega} f_m^*(t, \omega, \omega') Y_\nu(t, \omega') \mathbb{P}(d\omega');$$

- ▶ jump:

$$\begin{aligned} \Delta_\nu^s f^*(\omega) &\triangleq f(s, X^*(s, \omega), X^*(s) \# \mathbb{P}, \nu(\omega)) \\ &\quad - f(s, X^*(s, \omega), X^*(s) \# \mathbb{P}, u^*(s, \omega)), \end{aligned}$$



## Variational equation

$$\frac{d}{dt} Y_\nu(t, \omega) = f_x^*(t, \omega) \cdot Y_\nu(t, \omega) + \langle f_m^*, Y_\nu \rangle(\tau, \omega),$$
$$Y_\nu(s, \omega) = \Delta_\nu^s f^*(\omega).$$

## Variational equation

$$\begin{aligned}\frac{d}{dt} Y_\nu(t, \omega) &= f_x^*(t, \omega) \cdot Y_\nu(t, \omega) + \langle f_m^*, Y_\nu \rangle(\tau, \omega), \\ Y_\nu(s, \omega) &= \Delta_\nu^s f^*(\omega).\end{aligned}$$

Proposition (derivative of perturbed motions).

$$\frac{1}{h_n} \|Z_\nu^{h_n}(t) - X^*(t) - h_n Y_\nu(t)\|_{L^p} \rightarrow 0 \text{ as } n \rightarrow \infty$$

uniformly for  $t \in (s, T]$ .

## Variation of running cost. Notation

- ▶ derivative w.r.t.  $x$ :

$$f_{0,x}^*(t, \omega) \triangleq \nabla_x f_0(t, X^*(t, \omega), X^*(t) \# \mathbb{P}, u^*(t, \omega)),$$

- ▶ derivative w.r.t.  $m$ :

$$f_{0,m}^*(t, \omega, \omega') \triangleq \nabla_m f_0(t, X^*(t, \omega), X^*(t) \# \mathbb{P}, X^*(t, \omega'), u^*(t, \omega)),$$

- ▶ inner product:

$$\langle f_{0,m}^*, Y_\nu \rangle(t, \omega) \triangleq \int_{\Omega} f_{0,m}^*(t, \omega, \omega') Y_\nu(t, \omega') \mathbb{P}(d\omega');$$

- ▶ jump:

$$\begin{aligned} \Delta_\nu^s f_0^*(\omega) &\triangleq f(s, X^*(s, \omega), X^*(s) \# \mathbb{P}, \nu(\omega)) \\ &\quad - f_0(s, X^*(s, \omega), X^*(s) \# \mathbb{P}, u^*(s, \omega)). \end{aligned}$$

## Variation of the running cost

$$\begin{aligned} & \lim_{n \rightarrow \infty} \frac{1}{h_n} \left[ \int_0^T \mathbb{E} f_0(t, Z_\nu^{h_n}(t), Z_\nu^{h_n}(t) \# \mathbb{P}, u_\nu^{h_n}(t)) dt \right. \\ & \quad \left. - \int_0^T \mathbb{E} f_0(t, X^*(t), X^*(t) \# \mathbb{P}, u^*(t)) dt \right] \\ & = \mathbb{E} \Delta_\nu^s f_0^* + \int_s^T \mathbb{E} [f_{0,x}^*(t) Y_\nu(t) + \langle f_{0,m}^*(t), Y_\nu \rangle(t)] dt. \end{aligned}$$

## Variation of the terminal cost. Notation

- ▶ derivative w.r.t.  $x$ :

$$\sigma_x^*(\omega) \triangleq \nabla_x \sigma(X^*(T, \omega), X^*(T) \# \mathbb{P});$$

- ▶ derivative w.r.t.  $x$ :

$$\sigma_m^*(\omega, \omega') \triangleq \nabla_m \sigma(X^*(T, \omega), X^*(T) \# \mathbb{P}, X^*(T, \omega'));$$

- ▶ inner product:

$$\langle \sigma_m^*, Y_\nu \rangle(\omega) \triangleq \int_{\Omega} \sigma_m^*(\omega, \omega') Y_\nu(\omega') \mathbb{P}(d\omega').$$

## Variation of the terminal cost

$$\lim_{n \rightarrow \infty} \frac{1}{h_n} \mathbb{E} |\sigma(Z_\nu^{h_n}(T), Z_\nu^{h_n}(T) \# \mathbb{P}) - \sigma(X^*(T), X^*(T) \# \mathbb{P}) - h_n[\sigma_x^* + \langle \sigma_m^*, Y_\nu \rangle]| = 0.$$

Local minimizers and PMP for the Kantorovich approach

## Kantorovich approach

- ▶ **Control process:**  $(\eta, u_K)$ , where  $\eta \in \mathcal{P}^p(\Gamma)$ ,  $u_K \in L^p(\Gamma, \mathcal{B}(\Gamma), \eta; \mathcal{U}^p)$ .
- ▶ **Feasibility:** for  $\eta$ -a.e.  $\gamma \in \Gamma$ ,

$$\frac{d}{dt}\gamma(t) = f(t, \gamma(t), e_t \# \eta, u_K(t, \gamma)),$$

where  $e_t(\gamma) = \gamma(t)$ ,  $(e_t \# \eta)(E) = \eta\{\gamma \in \Gamma : \gamma(t) \in E\}$ .

- ▶ **Initial condition:**  $e_0 \# \eta = m_0$ .
- ▶ **Payoff:**

$$J_K(\eta, u_K) \triangleq \int_{\Gamma} \sigma(\gamma(T), e_T \# \eta) \eta(d\gamma) + \int_{\Gamma} \int_0^T f_0(t, \gamma(t), e_t \# \eta, u_K(t, \gamma)) dt \eta(d\gamma).$$



## Kantorovich local minimizer

A Kantorovich control process  $(\eta^*, u_K^*)$  is a **strong local minimizer** at  $m_0$  if there exists  $\varepsilon > 0$  such that: for any Kantorovich process  $(\eta, u_K)$  satisfying  $W_p(e_t \# \eta, e_t \# \eta^*) \leq \varepsilon$ ,

$$J_K(\eta^*, u_K^*) \leq J_K(\eta, u_K).$$

## Kantorovich and Lagrangian processes

Let

- ▶  $(\eta, u_K)$  be a Kantorovich control process;
- ▶  $(\Omega, \mathcal{F}, \mathbb{P})$  be a standard probability space.

A Lagrangian control process  $(X, u_L)$  defined on  $(\Omega, \mathcal{F}, \mathbb{P})$  **realizes**  $(\eta, u_K)$  if

- ▶  $\eta = X(\cdot) \# \mathbb{P}$ ,
- ▶  $u_L(t, \omega) = u_K(t, X(\cdot, \omega))$  for  $\mathbb{P}$ -a.e.  $\omega \in \Omega$  and a.e.  $t \in [0, T]$ .

# Kantorovich minimum implies Lagrangian minimum

## Theorem

Assume that  $(\eta^*, u_K^*)$  is a *strong local minimizer* in the framework of the Kantorovich approach at  $m_0 = e_t \# \eta^*$ . Let  $(X^*, u_L^*)$  be a Lagrangian process that *realizes* the Kantorovich process  $(\eta^*, u_K^*)$ . Then,  $(X^*, u_L^*)$  is a *strong local minimizer* at  $m_0$  in the framework of the Lagrangian approach.

# Kantorovich minimum implies Lagrangian minimum

## Theorem

*Let  $(\eta, u_K)$  be a Kantorovich control process. Assume also that the standard probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  is such that either the probability  $\mathbb{P}$  has no atoms or  $(\Omega, \mathcal{F}, \mathbb{P}) = (\Gamma, \mathcal{B}(\Gamma), \eta)$ .*

*Then, there exists a Lagrangian process  $(X, u_L)$  defined on  $(\Omega, \mathcal{F}, \mathbb{P})$  that **realizes**  $(\eta, u_K)$ .*

*Furthermore, if  $(\Omega, \mathcal{F}, \mathbb{P}) = (\Gamma, \mathcal{B}(\Gamma), \eta)$ , one can put  $X = \text{id}_\Omega$  and  $u_K = u_L$ .*

## Pontryagin maximum principle in the Kantorovich form

Let  $(\eta^*, u_K^*)$  be a strong local minimizer in the framework of the Kantorovich approach.

Then, there exists a function  $\psi \in L^q(\Gamma, \mathcal{B}(\Gamma), \eta^*; \Gamma^*)$  such that

- ▶ the costate equation,
- ▶ transversality condition,
- ▶ maximization conditions

hold true.

## Costate equation

For  $\eta^*$ -a.e.  $\gamma \in \Gamma$ ,  $\psi(\cdot, \gamma)$  solves

$$\begin{aligned} & \frac{d}{dt}\psi(t, \gamma) \\ &= -\psi(t, \gamma)\nabla_x f(t, \gamma(t), e_t \# \eta^*, u_K^*(t, \gamma)) \\ & \quad + \nabla_x f_0(t, \gamma(t), e_t \# \eta^*, u_K^*(t, \gamma)) \\ & \quad - \int_{\Gamma} \psi(t, \gamma')\nabla_m f(t, \gamma'(t), e_t \# \eta^*, \gamma(t), u_K^*(t, \gamma'))\eta^*(d\gamma') \\ & \quad + \int_{\Gamma} \nabla_m f_0(t, \gamma'(t), e_t \# \eta^*, \gamma(t), u_K^*(t, \gamma'))\eta^*(d\gamma'). \end{aligned}$$

## Transversality condition

$$\begin{aligned}\psi(T, \gamma) \# \eta^* &= -\nabla_x \sigma(\gamma(T), e_T \# \eta^*) \\ &\quad - \int_{\Gamma} \nabla_m \sigma(\gamma'(t), e_T \# \eta^*, \gamma(T)) \eta^*(d\gamma')\end{aligned}$$

for  $\eta^*$ -a.e.  $\gamma \in \Gamma$ .

## Maximization of the Hamiltonian condition

$$\begin{aligned} H(s, \gamma(s), e_s \# \eta^*, \psi(s, \gamma), u^*(s, \gamma)) \\ = \max_{u \in U} H(s, \gamma(s), e_s \# \eta^*, \psi(s, \gamma), u) \end{aligned}$$

for a.e.  $s \in [0, T]$  and  $\eta^*$ -a.e.  $\gamma \in \Gamma$ .



Local minimizers and PMP for the Eulerian approach

## Eulerian approach

- ▶ **Control process:**  $(m(\cdot), u_E)$ , where  $m(t)$  is a probability on  $\mathbb{R}^d$ ,  $u_E : [0, T] \times \mathbb{R}^d \rightarrow U$ .
- ▶ **Dynamics:**  $m(\cdot)$  is a distributional solution of the nonlocal continuity equation:

$$\partial_t m(t) + \operatorname{div}(v_E(t, x)m(t)) = 0,$$

for  $v_E(t, x) = f(t, x, m(t), u_E(t, x))$ .

- ▶ **Initial condition:**  $m(0) = m_0$ .
- ▶ **Payoff:**

$$J_E(\mu, u_E) \triangleq \int_{\mathbb{R}^d} \sigma(x, m(T))m(T, dx) + \int_0^T \int_{\mathbb{R}^d} f_0(t, x, m(t), u_E(t, x))m(t, dx)dt.$$

## Convexity condition for the Eulerian case

- ▶ the set  $U$  is a closed convex subset of a Banach space;
- ▶ the mapping  $U \ni u \mapsto f(t, x, m, u)$  is affine in  $u$ , i.e., for  $t \in [0, T]$ ,  $x \in \mathbb{R}^d$ ,  $m \in \mathcal{P}^p(\mathbb{R}^d)$ ,  $u_1, u_2 \in U$ ,  $\alpha \in [0, 1]$ ,

$$\begin{aligned} & f(t, x, m, \alpha u_1 + (1 - \alpha)u_2) \\ & = \alpha f(t, x, m, u_1) + (1 - \alpha)f(t, x, m, u_2); \end{aligned}$$

- ▶ the function  $f_0$  is convex in  $u$ , i.e., for every  $t \in [0, T]$ ,  $x \in \mathbb{R}^d$ ,  $m \in \mathcal{P}^p(\mathbb{R}^d)$ ,  $u_1, u_2 \in U$ ,  $\alpha \in [0, 1]$ ,

$$\begin{aligned} & f_0(t, x, m, \alpha u_1 + (1 - \alpha)u_2) \\ & \leq \alpha f_0(t, x, m, u_1) + (1 - \alpha)f_0(t, x, m, u_2). \end{aligned}$$

## Eulerian local minimizer

A Eulerian control process  $(m^*(\cdot), u_E^*)$  is a **strong local minimizer** at  $m_0$  if there exists  $\varepsilon > 0$  such that: for any Eulerian process  $(m(\cdot), u_E)$  satisfying  $W_p(m(t), m^*(t)) \leq \varepsilon$ ,

$$J_E(m^*(\cdot), u_E^*) \leq J_E(m(\cdot), u_E).$$

## Eulerian and Lagrangian processes

Let  $(m(\cdot), u_E)$  be an Eulerian control process. A Lagrangian control process  $(X, u_L)$  defined on a standard probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  realizes  $(m^*(\cdot), u_E)$  provided that

- ▶  $m(t) = X(t) \# \mathbb{P}$  for every  $t \in [0, T]$ ;
- ▶  $u_L(t, \omega) = u_E(t, X(t, \omega))$  for a.e.  $t \in [0, T]$  and  $\mathbb{P}$ -a.e.  $\omega \in \Omega$ .

# Eulerian minimizer implies Lagrangian one

## Theorem

Let  $(m^*(\cdot), u_E^*)$  be a *strong local minimizer* in the Euler framework and let  $(X^*, u_L^*)$  be a Lagrangian process defined on some standard probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  that *realizes*  $(m^*(\cdot), u_E^*)$ .

Then,  $(X^*, u_L^*)$  is a *strong local minimizer* at  $m_0$  within the Lagrangian framework.

## Eulerian minimizer implies Lagrangian one

### Theorem

Assume that  $(m(\cdot), u_E)$  is an Eulerian control process. Furthermore, let a standard probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  be such that at least one the following conditions satisfies:

- ▶ the probability  $\mathbb{P}$  has no atoms,
- ▶  $(\Omega, \mathcal{F}, \mathbb{P}) = (\Gamma, \mathcal{B}(\Gamma), \eta)$ , and  $\eta$  is concentrated on solutions of

$$\frac{d}{dt}\gamma(t) = f(t, \gamma(t), m(t), u_E(t, \gamma(t)))$$

and  $m(t) = e_t \# \eta$ .

Then, there exists a Lagrangian process  $(X, u_L)$  defined on  $(\Omega, \mathcal{F}, \mathbb{P})$  that **realizes**  $(m(\cdot), u_E)$ . Furthermore, in the second case, we can put  $X = \text{id}_\Omega$  and  $u_E(t, \gamma(t)) = u_L(t, \gamma)$ .

## Pontryagin maximum principle for the Eulerian framework

Let an Eulerian control process  $(m^*(\cdot), u_E^*)$  be a strong local minimizer at  $m_0$ . Then, there exists a flow of probabilities  $\nu^*(\cdot) : [0, T] \rightarrow \mathcal{P}^{p \wedge q}(\mathbb{R}^d \times \mathbb{R}^{d,*})$  satisfying

- ▶ consistency condition;
- ▶ continuity equation playing both the roles of state and costate equations;
- ▶ transversality condition;
- ▶ maximization of the Hamiltonian condition.



## Consistency condition

$$p^1 \# \nu^*(t) = m^*(t) \quad \forall t \in [0, T];$$

Here  $p^i(x_1, x_2) \triangleq x_i$ .

## Joint state and costate continuity equation

$\nu^*(\cdot)$  is a distributional solution of the continuity equation

$$\partial_t \nu^* + \operatorname{div}(\mathcal{J}(t, x, \psi) \nu^*) = 0,$$

where the vector field  $\mathcal{J}(t, x, \psi) = (\mathcal{J}_x(t, x, \psi), \mathcal{J}_\psi(t, x, \psi))$  is given by

$$\mathcal{J}_x(t, x, \psi) \triangleq f(t, x, m^*(t), u_E^*(t, x)),$$

$$\begin{aligned} \mathcal{J}_\psi(t, x, \psi) \triangleq & \\ & -\psi \nabla_x f(t, x, m^*(t), u_E^*(t, x)) \\ & + \nabla_x f_0(t, x, m^*(t), u_E^*(t, x)) \\ & - \int_{\mathbb{R}^d \times \mathbb{R}^{d,*}} \zeta \nabla_m f(t, y, m^*(t), x, u_E^*(t, y)) \nu^*(t, d(y, \zeta)) \\ & + \int_{\mathbb{R}^d} \nabla_m f_0(t, y, m^*(t), x, u_E^*(t, y)) m^*(t, dy); \end{aligned}$$

## Continuity equation

$\nu(\cdot)$  is a distributional solution of the nonlocal continuity equation:

$$\partial_t \nu(t) + \operatorname{div}(\mathcal{J}(t, x, \psi) \nu(t)) = 0,$$

iff, for every  $\varphi \in C_c^\infty((0, T) \times \mathbb{R}^d \times \mathbb{R}^{d,*})$ ,

$$\int_0^T \int_{\mathbb{R}^d \times \mathbb{R}^{d,*}} [\partial_t \varphi(t, x, \psi) + \nabla_x \varphi(t, x, \psi) \mathcal{J}_x(t, x, \psi) + \mathcal{J}_\psi(t, x, \psi) \nabla_\psi \varphi(t, x, \psi)] \nu(t, d(x, \psi)) dt = 0.$$

## Transversality condition

$$p^2 \# \nu^*(T) = \left[ -\nabla_x \sigma(\cdot, m^*(T)) - \int_{\mathbb{R}^d} \nabla_m \sigma(y, m^*(T), \cdot) m^*(T, dy) \right] \# m^*(T).$$

Here  $p^i(x_1, x_2) \triangleq x_i$ .

## Maximization condition

$$H(s, x, m^*(s), \psi, u_E^*(s, x)) = \max_{u \in U} H(s, x, m^*(s), \psi, u)$$

for almost every  $s \in [0, T]$  and  $\nu^*(s)$ -a.e.  $(x, \psi) \in \mathbb{R}^d \times \mathbb{R}^{d,*}$ .

## Costate equation in the Hamiltonian form

- ▶ Hamiltonian for the Eulerian framework:

$$\mathcal{H}(t, \nu, u) \triangleq \int_{\mathbb{R}^d \times \mathbb{R}^{d,*}} H(t, x, p^1 \# \nu, \psi, u) \nu(d(x, \psi)).$$

- ▶ Unit symplectic matrix:  $\mathbb{J}(\zeta, y) \triangleq (y, -\zeta)$ .
- ▶ Averaged terminal payoff:

$$\mathcal{S}(m) \triangleq \int_{\mathbb{R}^d} \sigma(x, m) m(dx).$$

State+costate equation:

$$\partial_t \nu^* + \operatorname{div}(\mathbb{J} \nabla_{\nu} \mathcal{H}(t, \nu^*(t), x, \psi, u_E^*(t, x)) \nu^*) = 0,$$

$$p^1 \# \nu^*(0) = m_0,$$

$$p^2 \# \nu^*(T) = -\nabla_m \mathcal{S}(p^1 \# \nu^*(T), \cdot) \# (p^1 \# \nu^*(T)).$$

Example. Mean field type linear-quadratic regulator

## Dynamics and payoffs

Dynamics of each agent:

$$\frac{d}{dt}X(t, \omega) = A(t)X(t, \omega) + B(t)u(t, \omega),$$

Total payoff:

$$\begin{aligned} \frac{1}{2} \mathbb{E} \left( \int_0^T [X^T(t) Q_x(t) X(t) + u^T(t) R(t) u(t)] dt \right. \\ \left. + X^T(T) K_x(t) X(T) \right) \\ + \frac{1}{2} \int_0^T \mathbb{E} [(X(t) - \mathbb{E}X(t))^T Q_m(t) (X(t) - \mathbb{E}X(t))] dt \\ + \frac{1}{2} \mathbb{E} [(X(T) - \mathbb{E}X(T))^T K_m (X(T) - \mathbb{E}X(T))]. \end{aligned}$$



## Optimal control

$$u^*(t, \omega) = -R^{-1}(t)B^T(t)[P_1(t)(X^*(t, \omega) - \mathbb{E}X^*(t)) + P_2(t)\mathbb{E}X^*(t)],$$

## Ricatti equation for $P_1$

$$\begin{aligned}\frac{d}{dt}P_1(t) = & -P_1(t)A(t) \\ & - A^T(t)P_1(t) \\ & + P_1(t)B(t)R^{-1}(t)B(t)P_1(t) \\ & - (Q_x(t) + Q_m(t));\end{aligned}$$

boundary condition:

$$P_1(T) = K_x + K_m,$$

## Ricatti equation for $P_2$

$$\begin{aligned} \frac{d}{dt}P_2(t) = & -P_2(t)A(t) - A^T(t)P_2(t) \\ & + P_2(t)B(t)R^{-1}(t)B(t)P_2(t) - Q_x(t); \end{aligned}$$

boundary condition:

$$P_2(T) = K_x.$$

Thank you for your attention!