# Extended Mean Field Control Problems with Singular Controls

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We consider the state dynamics

$$dX_t = b(t, m_t, X_t, \xi_t)dt + \sigma(t, m_t, X_t, \xi_t)dW_t + \gamma(t)d\xi_t,$$

where  $m_t := \mathbb{P}_{(X_t,\xi_t)}$  and  $\xi$  a non-decreasing, càdlàg control.

The goal is maximising the reward functional

$$J(\xi) := \int_0^T f(t, m_t) dt + g(m_T) - \mathbb{E}\left[\int_{[0,T]} c(t) d\xi_t\right]$$

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 $\rightarrow$  How to extend this to  $c(t, m_t, X_t, \xi_t)$ ?

#### Connection to Regular Controls

Suppose  $\xi$  is absolutely continuous with  $\|\dot{\xi}_t\| \leq K$ .  $\rightarrow$  bounded velocity We can view instead  $u_t := \dot{\xi}_t$  as the control for the new dynamics

$$dX_t = [b(t, X_t, \xi_t) + \gamma(t)u_t] dt + \sigma(t, X_t, \xi_t) dW_t$$
  
$$d\xi_t = u_t dt.$$

The reward functional can then be written as

$$J(u) = \mathbb{E}\left[\int_0^T f(t, X_t, \xi_t) dt + g(X_T, \xi_T) - \int_0^T c(t) \cdot u_t dt\right].$$

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 $\rightarrow$  Singular controls arise as the limit of bounded velocity controls.

#### The Extended Reward Functional

We define

$$J(t,m;\xi) := \int_t^T f(s,m_s) ds + g(m_T) - \mathbb{E}\left[\int_t^T c(s,m_s,X_s,\xi_s) d\xi_s\right],$$

for absolutely continuous controls.

For general singular controls we define the reward as the maximal reward we can get via absolutely continuous approximations

$$J(t, m, \xi) \coloneqq \sup_{\xi^n \to \xi} \limsup_{n \to \infty} J(t, m, \xi^n).$$

The value function is defined as usual

$$V(t,m) := \sup_{\xi, \xi_{t-}=m} J(t,m,\xi).$$

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ightarrow Use the weak  $M_1$  topology.

The  $M_1$  topology in  $\mathbb{R}$ 







So we define the costs for such a jump from  $(t-,\xi_{t-})$  to  $(t,\xi_t)$  as

$$\int_0^{\xi_t-\xi_{t-}} c(t,m_t,X_{t-}+\gamma(t)\zeta,\xi_{t-}+\zeta)d\zeta.$$





The Weak  $M_1$  topology in  $\mathbb{R}^d$ 



So we define

$$C_{\xi}(t,m,X_{t-},\xi_{t-},\xi_t) \coloneqq \inf_{\zeta \in \Xi(\xi,\xi')} \int_0^1 c(t,m,X_{t-}+\gamma(t)(\zeta_{\lambda}-\xi_{t-}),\zeta_{\lambda}) d\zeta_{\lambda},$$

over the set  $\Xi(\xi,\xi')$  of all absolutely continuous and monotone paths from  $\xi_{t-}$  to  $\xi_t$ .

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ightarrow We can generalise this idea to jumps in  $t\mapsto m_t\in\mathcal{P}_2$  and define

 $C_m(t, m, m')$ 

as the minimal costs of an interpolating path from (t-, m) to (t, m').

## An Explicit Representation of J

#### Theorem

The reward functional J defined before has the following alternative characterisation

$$\begin{aligned} J(t, m, \xi) \\ &= \int_{t}^{T} f(s, m_{s}) ds + g(m_{T}) - \sum_{J_{[t, T]}(m)} C_{m}(s, m_{s-}, m_{s}) \\ &- \mathbb{E} \bigg[ \sum_{J_{[t, T]}^{c}(m) \cap J_{[t, T]}(\xi)} C_{\xi}(s, m_{s}, X_{s-}, \xi_{s-}, \xi_{s}) \\ &+ \int_{J_{[t, T]}^{c}(m) \cap J_{[t, T]}^{c}(\xi)} c(s, m_{s}, X_{s}, \xi_{s}) d\xi_{s} \bigg], \end{aligned}$$

where J and J<sup>c</sup> denote the jump and continuity sets respectively.

#### Dynamic Programming Principle

#### Theorem

Let  $(t, m) \in [0, T] \times \mathcal{P}_2(\mathbb{R}^d \times \mathbb{R}^l)$ . For all  $s \in [t, T]$ , we have the following dynamic programming principle

$$V(t, m) = \sup_{\xi, \xi_{t-}=m} \left[ V(s, m_{s-}) + \int_{t}^{s} f(r, m_{r}) dr - \sum_{J_{[t,s]}(m)} C_{m}(r, m_{r-}, m_{r}) \right. \\ \left. - \mathbb{E} \left[ \sum_{J_{[t,s]}^{c}(m) \cap J_{[t,s]}(\xi)} C_{\xi}(r, m_{r}, X_{r-}, \xi_{r-}, \xi_{r}) \right. \\ \left. + \int_{J_{[t,s]}^{c}(m) \cap J_{[t,s]}^{c}(\xi)} c(r, m_{r}, X_{r}, \xi_{r}) d\xi_{r} \right] \right].$$

#### Wasserstein Calculus

A function  $u: \mathcal{P}_2(\mathbb{R}^d) \to \mathbb{R}$  admits a linear derivative if there exists a function  $\delta_m u: \mathcal{P}_2(\mathbb{R}^d) \times \mathbb{R}^d \to \mathbb{R}$  such that

$$u(m')-u(m)=\int_0^1\int_{\mathbb{R}^d}\delta_m u(\lambda m'+(1-\lambda)m,x)(m'-m)(dx)d\lambda.$$

As example, for functions of the form

$$u(m) = \int_{\mathbb{R}^d \times \mathbb{R}^l} \psi(x) m(dx),$$

the linear derivative is given, up to an additive constant, by

$$\delta_m u(m,x) = \psi(x).$$

### QVI, Continuation Region

Theorem (Itô formula without jumps by Cosso et al, 2022) Let  $u \in C_2^{1,2}$  and  $\xi_r = \xi_{t-}$  for all  $r \in [t, s]$ . Then

$$u(s,m_s)=u(t,m_t)+\int_t^s\mathbb{L}u(r,m_r)dr,$$

where  $\mathbb{L}$  is the infinitesimal generator of  $(t, X_t)_{t \in [0,T]}$  defined by

$$\begin{aligned} \mathbb{L}u(t,m) &\coloneqq \partial_t u(t,m) + \int_{\mathbb{R}^d \times \mathbb{R}^d} b(t,m,x,\xi) \cdot \partial_x \delta_m u(t,m,x,\xi) \\ &+ \frac{1}{2} \sigma \sigma^T(t,m,x,\xi) : \partial_{xx}^2 \delta_m u(t,m,x,\xi) m(dx,d\xi). \end{aligned}$$

 $\rightarrow$  In the continuation region the optimal control is  $\xi_t = \xi_{t-}$ , hence we can derive an HJB from the DPP and Itô's formula.

$$-\mathbb{L}u-f=0.$$

### QVI, Intervention Region

The intervention region should be characterised by

$$V(t,m) = \sup_{m' \leq t, m, m' \neq m} [V(t,m') - C_m(t,m,m')],$$

where  $\{m' \leq_t m\}$  are the reachable states (t, m') starting from (t-, m).

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#### Lemma

Assume that  $u \in C_2^{1,2}$  and let  $t \in [0, T]$  be fixed. Let  $m \in \mathcal{P}_2(\mathbb{R}^d \times \mathbb{R}^l)$ , if we have

$$u(t,m)+C_m(t,m,m')\geq u(t,m')$$
 for all  $m'\preceq_t m,$ 

then

$$\partial_{(x,\xi)}\delta_m u(t,m,x,\xi) \cdot (\gamma(t),1) \leq c(t,m,x,\xi) \qquad \text{for all } (x,\xi) \in \mathbb{R}^d imes \mathbb{R}^l.$$

#### Quasi-Variational Inequality

#### Theorem

Suppose the value function V is continuous, then it is the minimal bounded viscosity supersolution to

$$\begin{split} \min \left\{ & -\mathbb{L}u(t,m) - f(t,m), \\ & \inf_{\substack{(x,\xi) \in \mathbb{R}^d \times \mathbb{R}^d}} \left[ c(t,m,x,\xi) - \partial_{(x,\xi)} \delta_m u(t,m,x,\xi) \cdot (\gamma(t),1) \right] \right\} = 0, \\ \min \left\{ u(T,m) - g(T,m), \\ & \inf_{\substack{(x,\xi) \in \mathbb{R}^d \times \mathbb{R}^d}} \left[ c(T,m,x,\xi) - \partial_{(x,\xi)} \delta_m u(T,m,x,\xi) \cdot (\gamma(t),1) \right] \right\} = 0. \end{split}$$

## Proof Idea: Connection to Regular Controls (revisited)

Recall the bounded velocity control problem: We consider a regular control u with  $u \ge 0$  and  $||u|| \le K$  and

$$dX_t = [b(t, m_t, X_t, \xi_t) + \gamma(t)u_t] dt + \sigma(t, m_t, X_t, \xi_t) dW_t$$
  
$$d\xi_t = u_t dt.$$

We want to maximise

$$J(t,m;u) = \int_t^T f(t,m_t)dt + g(m_T) - \mathbb{E}\left[\int_{[t,T]} c(t,m_t,X_t,\xi_t) \cdot u_t dt\right]$$

Define the corresponding value function as

$$V_{K}(t,m) := \sup_{u \ge 0, \|u\| \le K} J(t,m;u).$$

 $\rightarrow$  We expect  $V_{\mathcal{K}} \uparrow V$  as  $\mathcal{K} \rightarrow \infty$ .

### Proof Idea: The Master Equation for $V_K$

The master equation for the bounded velocity problem is

$$\begin{split} -\mathbb{L}\mathbf{v}(t,m) - f(t,m) \\ + \int_{\mathbb{R}^d \times \mathbb{R}^l} \inf_{u \ge 0, \|u\| \le \kappa} \left\{ \left[ c(t,m,x,\xi) - \partial_{(x,\xi)} \delta_m \mathbf{v}(t,m,x,\xi) \cdot (\gamma(t),1) \right] \cdot u \right\} m(dx,d\xi) \\ &= 0, \\ \mathbf{v}(T,\cdot) = g. \end{split}$$

 $\rightarrow$  This equation looks very similar to our QVI.

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# Thank you

#### preprint on arXiv soon