

# Mean-Field optimization problems and non-anticipative optimal transport

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Robust Methods in Probability & Finance  
ICERM, Brown University, June 19-23, 2017

# Outline

- 1 McKean-Vlasov control problem and motivation
- 2 Our toolkit: causal transport
- 3 Characterization of MKV solutions via causal transport
- 4 Conclusions and ongoing research

# N-player stochastic differential game

→  $N$  players with **private state processes** given by the solutions to

$$dX_t^{N,i} = b_t(X_t^{N,i}, \alpha_t^{N,i}, \bar{v}_t^{N,i})dt + dW_t^i, \quad i = 1, \dots, N$$

- $W^1, \dots, W^N$  independent Wiener processes
- $\alpha^{N,1}, \dots, \alpha^{N,N}$  controls of the  $N$  players
- $\bar{v}_t^{N,i} = \frac{1}{N-1} \sum_{j \neq i} \delta_{X_t^{N,j}}$  empirical distrib. states of the other players

→ The **objective** of player  $i$  is to choose a control  $\alpha^{N,i}$  that minimizes

$$\mathbb{E} \left[ \int_0^T f_t(X_t^{N,i}, \alpha_t^{N,i}, \bar{\eta}_t^{N,i}) dt + g(X_T^{N,i}, \bar{v}_T^{N,i}) \right]$$

- $\bar{\eta}_t^{N,i} = \frac{1}{N-1} \sum_{j \neq i} \delta_{(X_t^{N,j}, \alpha_t^{N,j})}$  empirical joint distrib. of states & controls

→ Statistically identical players: same functions  $b_t, f_t, g$

# From N-player game to McKean-Vlasov control problem

- rarely expect existence of global minimizers
- resort to approximation by **asymptotic arguments**:

SDE State Dynamics  
for N players

optimization  
→

Nash equilibrium  
for N players

lim ↓  
N→∞

↓ lim  
N→∞

Mean-Field Game

McKean-Vlasov dynamics

optimization  
→

controlled McK-V dyn

Vast literature: Caines, Carmona, Delarue, Huang, Lachapelle, Lacker, Lasry, Lions, Malhamé, Pham, Sznitman, Wei,...

# The red path: approximating cooperative equilibria

## Main idea:

- all agents adopt the **same feedback control**:  $\alpha_t^{N,i} = \phi(t, X_t^{N,i})$
- in the limit ( $\#$  players  $\rightarrow \infty$ ) the private states of players evolve independently of each other
- distribution of private state converges toward distribution of the solution to the **McKean-Vlasov control problem**:

$$\inf_{\alpha} \mathbb{E} \left[ \int_0^T f_t(X_t, \alpha_t, \mathcal{L}(X_t, \alpha_t)) dt + g(X_T, \mathcal{L}(X_T)) \right]$$

$$\text{subject to } dX_t = b_t(X_t, \alpha_t, \mathcal{L}(X_t)) dt + dW_t$$

- under suitable conditions, the optimal feedback controls are  **$\epsilon$ -optimal** for large systems of players

# The blue path: approximating competitive equilibria

## Main idea:

- Seek for **Nash equilibria** for the  $N$ -player game
- Model behaviour of a representative agent, and solve the **Mean-Field Game problem**:

1) for every fixed joint law  $\eta$ , with first marginal  $\nu$ , solve

$$\inf_{\alpha} \mathbb{E} \left[ \int_0^T f_t(X_t, \alpha_t, \eta_t) dt + g(X_T, \nu_T) \right]$$

s.t.  $dX_t = b_t(X_t, \alpha_t, \nu_t) dt + dW_t$

2) fixed point problem:  $\eta$  s.t. for the solution  $\mathcal{L}(X, \alpha) = \eta$

- under suitable conditions, the optimal feedback provides an **approximate Nash equilibrium** for large system of players
- for **potential games**, MFG can be formulated as MKV

# McKean-Vlasov control problem

→ As a result of either approximation path, we shall study the following **McKean-Vlasov control problem**:

$$\inf_{\alpha} \mathbb{E} \left[ \int_0^T f_t(X_t, \alpha_t, \mathcal{L}(X_t, \alpha_t)) dt + g(X_T, \mathcal{L}(X_T)) \right]$$

subject to

$$dX_t = b_t(X_t, \alpha_t, \mathcal{L}(X_t)) dt + dW_t, \quad X_0 = 0$$

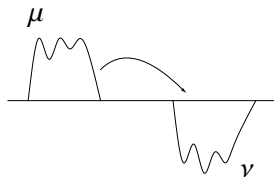
**Classical approaches:**

- **HJB/PDE** (Lasry-Lions): forward-backward system of PDEs
- **probabilistic**: Pontryagin maximum principle, adjoint FBSDEs

**Our approach:** use **Optimal Transport theory**

# Classical Monge-Kantorovich optimal transport

Given two Polish probability spaces  $(\mathcal{X}, \mu)$ ,  $(\mathcal{Y}, \nu)$ , move the mass from  $\mu$  to  $\nu$  **minimizing the cost of transportation**  $c : \mathcal{X} \times \mathcal{Y} \rightarrow [0, \infty]$



$$\text{OT}(\mu, \nu, c) := \inf \{ \mathbb{E}^\pi [c(x, y)] : \pi \in \Pi(\mu, \nu) \},$$

$\Pi(\mu, \nu)$ : probability measures on  $\mathcal{X} \times \mathcal{Y}$  with marginals  $\mu$  and  $\nu$ .

**Monge transport:** all mass sitting on  $x$  is transported into  $y = T(x)$ .

**Kantorovich transport:** mass can split.



# Causal( $\equiv$ non-anticipative) optimal transport

**Idea:** introduce **time**, and move the mass in a **non-anticipative** way: what is transported into the  $2^{nd}$  coordinate at time  $t$ , depends on the  $1^{st}$  coordinate only up to  $t$  (+ sth independent)

Let  $\mathcal{F}^X = (\mathcal{F}_t^X)_t$  on  $\mathcal{X}$ ,  $\mathcal{F}^Y = (\mathcal{F}_t^Y)_t$  on  $\mathcal{Y}$  be right-cont. filtrations.

**Definition ( Causal transport plans  $\Pi_c(\mu, \nu)$  )**

A **transport plan**  $\pi \in \Pi(\mu, \nu)$  is called **causal** between  $(\mathcal{X}, \mathcal{F}^X, \mu)$  and  $(\mathcal{Y}, \mathcal{F}^Y, \nu)$  if, for all  $t$  and  $D \in \mathcal{F}_t^Y$ , the map  $\mathcal{X} \ni x \mapsto \pi^x(D)$  is measurable w.t.to  $\mathcal{F}_t^X$  ( $\pi^x$  regular conditional kernel w.r.to  $\mathcal{X}$ ).

The concept goes back to Yamada-Watanabe (1971); see also Jacod (1980), Kurtz (2014), Lassalle (2015), Backhoff et al. (2016).

$$\text{COT}(\mu, \nu, c) := \inf \{ \mathbb{E}^\pi [c(X, Y)] : \pi \in \Pi_c(\mu, \nu) \}$$

## Example: weak-solutions of SDEs

- $\mathcal{X} = \mathcal{Y} = C_0[0, \infty)$
- $\mathcal{F}$  right-continuous canonical filtration

### Example (Yamada-Watanabe'71)

Assume weak-existence of the solution to the SDE:

$$dY_t = b(Y_t)dt + \sigma(Y_t)dB_t, \quad b, \sigma \text{ Borel measurable.}$$

Then  $\mathcal{L}(B, Y)$  is a causal transport plan between the spaces  $(C_0[0, \infty), \mathcal{F}, \mathcal{L}(B))$  and  $(C_0[0, \infty), \mathcal{F}, \mathcal{L}(Y))$ .

- **Transport perspective:** from an observed trajectory of  $B$ , the mass can be split at each moment of time into  $Y$  only based on the information available up to that time.
- **No splitting of mass:**

**Monge transport**  $\iff$  **strong solution**  $Y = F(B)$ .

# McKean-Vlasov control problem and Causal Transport

→ Recall our McKean-Vlasov control problem:

$$\inf_{\alpha} \mathbb{E} \left[ \int_0^T f_t(X_t, \alpha_t, \mathcal{L}(X_t, \alpha_t)) dt + g(X_T, \mathcal{L}(X_T)) \right]$$

subject to

$$dX_t = b_t(X_t, \alpha_t, \mathcal{L}(X_t)) dt + dW_t, \quad X_0 = 0$$

→ The joint distribution  $\mathcal{L}(W, X)$  is a causal transport plan between  $(C_0[0, T], \mathcal{F}, \mathcal{L}(W))$  and  $(C_0[0, T], \mathcal{F}, \mathcal{L}(X))$

# McKean-Vlasov control problem

**Definition.** A weak solution to the McKean-Vlasov control problem is a tuple  $(\Omega, (\mathcal{F}_t)_{t \in [0, T]}, \mathbb{P}, W, X, \alpha)$  such that:

- (i)  $(\Omega, (\mathcal{F}_t)_{t \in [0, T]}, \mathbb{P})$  supports  $X$ , BM  $W$ ,  $\alpha$  is  $\mathcal{F}$ -progress. meas.
- (ii) the state equation  $dX_t = b_t(X_t, \alpha_t, \mathbb{P} \circ X_t^{-1}) dt + dW_t$  holds
- (iii) if  $(\Omega', (\mathcal{F}'_t)_{t \in [0, T]}, \mathbb{P}', W', X', \alpha')$  is another tuple s.t. (i)-(ii) hold,

$$\begin{aligned} & \mathbb{E}^{\mathbb{P}} \left[ \int_0^T f_t(X_t, \alpha_t, \mathbb{P} \circ (X_t, \alpha_t)^{-1}) dt + g(X_T, \mathbb{P} \circ X_T^{-1}) \right] \\ & \leq \mathbb{E}^{\mathbb{P}'} \left[ \int_0^T f_t(X'_t, \alpha'_t, \mathbb{P}' \circ (X'_t, \alpha'_t)^{-1}) dt + g(X'_T, \mathbb{P}' \circ X'_T^{-1}) \right] \end{aligned}$$

# Assumptions

→ We need some **convexity assumptions**.

→ In the **case of linear drift**:

$$dX_t = (c_t^1 X_t + c_t^2 \alpha_t + c_t^3 \mathbb{E}[X_t])dt + dW_t,$$

$c_t^i \in \mathbb{R}$ ,  $c_t^2 > 0$ , the assumptions reduce to: for all  $x, a, \eta$ ,

- $f_t$  is bounded from below uniformly in  $t$
- $f_t(x, \cdot, \eta)$  is convex
- $f_t(x, a, \cdot)$  is  $<_{\text{conv}}$ -monotone

## Example: Inter-bank systemic risk model

[Carmona-Fouque-Sun 2013]

- Inter-bank borrowing/lending, where the **log-monetary reserve** of each bank, asymptotically, is governed by the MKV eq.

$$dX_t = [k(\mathbb{E}[X_t] - X_t) + \alpha_t]dt + dW_t, X_0 = 0$$

$k \geq 0$  rate of m-r in the interaction from b&l between banks

- All banks can control their rate of borrowing/lending to a central bank with the same policy  $\alpha$ , to **minimize the cost**

$$\mathbb{E} \left[ \int_0^T \left( \frac{1}{2} \alpha_t^2 - q \alpha_t (\mathbb{E}[X_t] - X_t) + \frac{c}{2} (\mathbb{E}[X_t] - X_t)^2 \right) dt + \frac{d}{2} (\mathbb{E}[X_T] - X_T)^2 \right]$$

$q > 0$  incentive to borrowing ( $\alpha_t > 0$ ) or lending ( $\alpha_t < 0$ ),  
 $c, d > 0$  penalize departure from average

# Characterization via non-anticipative optimal transport

- we use transport problems in the path space  $C := C_0[0, T]$
- $\gamma$ : Wiener measure on  $C$ ,  $(\omega, \bar{\omega})$ : generic element on  $C \times C$
- here for simplicity control = drift

## Theorem

Under the above assumptions, the *weak MKV* problem is **equivalent** to the *variational problem*

$$\inf_{v \in \tilde{\mathcal{P}}} \inf_{\pi \in \Pi_c(\gamma, v)} \mathbb{E}^\pi \left[ \int_0^T f_t(\bar{\omega}_t, (\overline{\dot{\omega}} - \dot{\omega})_t, p_t((\bar{\omega}, \overline{\dot{\omega}} - \dot{\omega})_{\#} \pi)) dt + g(\bar{\omega}_T, v_T) \right]$$

where  $p_t(\eta) = \eta_t$  for  $\eta \in \mathcal{P}(C)$ , and

$$\tilde{\mathcal{P}} = \{v \in \mathcal{P}(C) : v\text{-a.s. pathwise quadr.var. } \exists \text{ and } \langle \omega \rangle_t = t \forall t\}$$

# Characterization via non-anticipative optimal transport

'**Equivalence**' means that the above variational problem and

$$\inf \mathbb{E}^{\mathbb{P}} \left[ \int_0^T f_t (X_t, \alpha_t, \mathbb{P} \circ (X_t, \alpha_t)^{-1}) dt + g (X_T, \mathbb{P} \circ X_T^{-1}) \right]$$

have the **same value**, where the infimum is taken over tuples  $(\Omega, (\mathcal{F}_t), \mathbb{P}, W, X, \alpha)$  s.t.  $dX_t = b_t (X_t, \alpha_t, \mathbb{P} \circ X_t^{-1}) dt + dW_t$ , and that moreover the optimizers are related via:

- $\nu^* = \mathcal{L}(X^*)$
- $\pi^* \longleftrightarrow \alpha^*$ , with  $\pi = \mathcal{L}(W^*, X^*)$



# Characterization via non-anticipative optimal transport

→ Weak solutions of MKV control problem given by infimum over tuples  $(\Omega, (\mathcal{F}_t)_{t \in [0, T]}, \mathbb{P}, W, X, \alpha)$ .

## Corollary

- 1 The infimum can be taken over tuples s.t.  $\alpha$  is  $\mathcal{F}^X$ -measurable (*weak closed loop*).
- 2 If the infimum is *attained*, then the optimal  $\alpha$  is of weak closed loop form.

**Remark.** The outer minimization in VP can be done over  $\{\nu \ll \gamma\}$  instead of  $\tilde{\mathcal{P}}$ , whenever the drift is guaranteed to be square integr. (e.g. drift = control, and  $f_t(x, a, \eta) \geq K|a|^2 \quad \forall x, \eta$  and for a large).

# Example: separable cost

**Separable cost:** when running cost =  $f_t(x, a) + \tilde{f}_t(v_t, x)$ ,

$$\inf_{v \in \tilde{\mathcal{P}}} \left\{ \text{COT}(\gamma, v, c(f)) + P_{\tilde{f}}(v) \right\}, \quad P_{\tilde{f}}(v) \text{ penalty term}$$

↑  
standard causal transport (A.-Backhoff-Zalashko 2016)

**Example:** take  $k = q = 0$  in the example above, then

- state dynamics:  $dX_t = \alpha_t dt + dW_t$

- cost:  $\mathbb{E} \left[ \int_0^T \left( \frac{1}{2} \alpha_t^2 + \frac{c}{2} (\mathbb{E}[X_t] - X_t)^2 \right) dt + \frac{d}{2} (\mathbb{E}[X_T] - X_T)^2 \right]$

⇒ COT w.r.t. Cameron-Martin distance (Lassalle 2015):

$$\inf_{\pi \in \Pi_c(\gamma, \nu)} \mathbb{E}^\pi [|\bar{\omega} - \omega|_H^2] = \mathcal{H}(v|\gamma),$$

thus

$$\inf_{v \ll \gamma} \left\{ \mathcal{H}(v|\gamma) + \frac{c}{2} \int_0^T \text{Var}(v_t) dt + \frac{d}{2} \text{Var}(v_T) \right\}$$

# Conclusions

## Done so far:

- **connection** of McKean-Vlasov control problems to non-anticipative transport problems
- **characterization** of weak McKean-Vlasov solutions via non-anticipative transport

## Work in progress:

- The optimization over  $\Pi_c(\gamma, \nu)$  is not a standard causal transport problem  $\Rightarrow$  new analysis for **existence/duality**
- Use our characterization theorem in order to find
  - ▶ **existence and uniqueness** of weak MKV solutions
  - ▶ **explicit formulation** of solutions to MKV control problems
- Time-discretization and numerical scheme

**Thank you for your attention!**