Lagrangian, Eulerian and Kantorovich formulations of multi-agent optimal control problems



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Consider the evolution of a vector \mathbf{X}_t of N particles/agents indexed by $\omega = 1, 2, \cdots, N$:

$$\mathbf{X}_{t} = (X_{t}^{1}, X_{t}^{2}, \cdots, X_{t}^{N}) \rightsquigarrow (X_{t}(1), X_{t}(2), \cdots X_{t}(N)) = (X_{t}(\omega))_{\omega=1,2,\cdots,N}$$

 \mathbf{X} solves a system of ODEs

 $\frac{d}{dt}X_t(\omega) = F(X_t(\omega), X_t, U_t(\omega)) \quad t \in (0, T), \ X_{t=0}(\omega) = X_0(\omega), \qquad \omega \in \Omega^N = \{1, 2, \cdots, N\},$

driven by the vector field

$$F = F(\mathbf{x}, \mathbf{x}, \mathbf{u}) : \mathbb{R}^d \times (\mathbb{R}^d)^N \times \mathbb{U} \to \mathbb{R}^d$$

depending on

- the position $x = X_t(\omega)$ of each particle,
- the vector $\mathbf{x} = \mathbf{X}_{t}$ of the distribution of the particles
- the control variable $u = U_t(\omega)$ for each particle, varying in the set of controls U.

Ciao

Consider the evolution of a vector \boldsymbol{X}_t of N particles/agents indexed by $\boldsymbol{\omega}=1,2,\cdots$, N:

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driven by the vector field

$$\mathsf{F} = \mathsf{F}(\mathsf{x}, \mathbf{x}, \mathbf{u}) : \mathbb{R}^{\mathsf{d}} \times (\mathbb{R}^{\mathsf{d}})^{\mathsf{N}} \times \mathbb{U} \to \mathbb{R}^{\mathsf{d}}$$

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Consider the evolution of a vector X_t of N particles/agents indexed by $\omega = 1, 2, \cdots, N$:

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$$\frac{d}{dt}X_t(\omega) = F(X_t(\omega), \mathbf{X}_t, \mathbf{U}_t(\omega)) \qquad t \in (0, T), \ X_{t=0}(\omega) = X_0(\omega), \qquad \omega \in \Omega^N = \{1, 2, \cdots, N\},$$



Examples

First order drift-interaction systems:

$$F(x, \mathbf{x}, \mathbf{u}) = f(x, \mathbf{u}) + \frac{1}{N} \sum_{\omega=1}^{N} g(x - \mathbf{x}(\omega))$$

Linear controls: $f(x, \mathbf{u}) := f(x) + L\mathbf{u}$.

Second order systems: $\mathbf{x} \rightsquigarrow (\mathbf{x}, \mathbf{v}) \in \mathbb{R}^{2d}, \mathbf{x} \rightsquigarrow (\mathbf{x}, \mathbf{v}) \in (\mathbb{R}^{2d})^{\mathbf{N}}$

 $F(\mathbf{x}, \mathbf{v}, \mathbf{x}, \mathbf{v}, \mathbf{u}) = \left(\mathbf{v}, f(\mathbf{x}) + \frac{1}{N} \sum_{\omega=1}^{N} g(\mathbf{x} - \mathbf{x}(\omega)) + \mathbf{u}\right)$ corresponding to the system for the particles $(X_t(\omega), V_t(\omega))$ in the phase space

$$\begin{cases} \frac{d}{dt} X_t(\omega) = V_t(\omega), \\ \\ \frac{d}{dt} V_t(\omega) = f(X_t(\omega)) + \frac{1}{N} \sum_{\theta=1}^N g(X_t(\omega) - X_t(\theta)) + U_t(\omega) \end{cases}$$

Examples

First order drift-interaction systems:

$$F(x, \mathbf{x}, \mathbf{u}) = f(x, \mathbf{u}) + \frac{1}{N} \sum_{\omega=1}^{N} g(x - \mathbf{x}(\omega))$$

Linear controls: $f(x, \mathbf{u}) := f(x) + L\mathbf{u}$.

Second order systems:
$$x \rightsquigarrow (x, v) \in \mathbb{R}^{2d}, x \rightsquigarrow (x, v) \in (\mathbb{R}^{2d})^{N}$$

$$\begin{split} F(x,\nu,\boldsymbol{x},\boldsymbol{\nu},\boldsymbol{u}) &= \left(\nu,f(x)+\frac{1}{N}\sum_{\omega=1}^{N}g(x-\boldsymbol{x}(\omega))+\boldsymbol{u}\right) \text{ corresponding to the system for the particles } (X_t(\omega),V_t(\omega)) \text{ in the phase space} \end{split}$$

$$\begin{cases} \frac{d}{dt} X_t(\omega) = V_t(\omega), \\ \\ \frac{d}{dt} V_t(\omega) = f(X_t(\omega)) + \frac{1}{N} \sum_{\theta=1}^{N} g(X_t(\omega) - X_t(\theta)) + \frac{U_t(\omega)}{N} \\ \end{cases}$$

We introduce the running and final cost functions

 $\mathbf{C} = \mathbf{C}(\mathbf{x}, \mathbf{x}, \mathbf{u}) : \mathbb{R}^d \times (\mathbb{R}^d)^{\mathbf{N}} \times \mathbb{U} \to \mathbb{R}, \qquad \mathbf{C}_{\mathsf{T}} = \mathbf{C}_{\mathsf{T}}(\mathbf{x}, \mathbf{x}) : \mathbb{R}^d \times (\mathbb{R}^d)^{\mathbf{N}} \to \mathbb{R}$

and we want to minimize the total cost

$$\mathcal{J}(\boldsymbol{X},\boldsymbol{U}) := \frac{1}{N} \sum_{\omega=1}^{N} \int_{0}^{T} \boldsymbol{C}(\boldsymbol{X}_{t}(\omega),\boldsymbol{X}_{t},\boldsymbol{U}_{t}(\omega)) \, dt + \frac{1}{N} \sum_{\omega=1}^{N} \boldsymbol{C}(\boldsymbol{X}_{T}(\omega),\boldsymbol{X}_{T}).$$

among all the admissible pairs $\,(X,\,{\color{black}U})\in\mathcal{A}^{\,N}(X_{0})$ solving

 $\frac{d}{dt}X_t(\omega)=F(X_t(\omega),\boldsymbol{X}_t,\boldsymbol{U}_t(\omega)),\quad t\in(0,T),\quad X_{t=0}(\omega)=X_0(\omega).$

 $\begin{array}{ll} \mbox{Value function:} & \mathcal{V}^N(X_0):=\inf\left\{ \mathfrak{J}(X,\mathbf{U}):(X,\mathbf{U})\in\mathcal{A}^N(X_0)\right\}\\ \\ \mbox{Example:} & \mathbb{U}:=\mathbb{R}^d, \quad \mathbf{C}(\mathbf{x},\mathbf{x},\mathbf{u}):=V(\mathbf{x})+\frac{1}{N}\sum_{m=1}^N W(\mathbf{x}-\mathbf{x}(\omega))+\psi(\mathbf{u}) \end{array}$

We introduce the running and final cost functions

 $\mathbf{C} = \mathbf{C}(\mathbf{x}, \mathbf{x}, \mathbf{u}) : \mathbb{R}^{d} \times (\mathbb{R}^{d})^{\mathbf{N}} \times \mathbb{U} \to \mathbb{R}, \qquad \mathbf{C}_{\mathsf{T}} = \mathbf{C}_{\mathsf{T}}(\mathbf{x}, \mathbf{x}) : \mathbb{R}^{d} \times (\mathbb{R}^{d})^{\mathbf{N}} \to \mathbb{R}$

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among all the admissible pairs $\,(X,\,{\color{black}U})\in\mathcal{A}^{\,N}(X_{0})$ solving

$$\frac{d}{dt}X_t(\omega) = F(X_t(\omega), \boldsymbol{X}_t, \boldsymbol{U}_t(\omega)), \quad t \in (0, T), \quad X_{t=0}(\omega) = X_0(\omega).$$

 $\textbf{Value function:}\qquad \mathcal{V}^{N}(\textbf{X}_{0}):=\inf\left\{ \mathcal{J}(\textbf{X},\textbf{U}):(\textbf{X},\textbf{U})\in\mathcal{A}^{N}(\textbf{X}_{0})\right\}$

Example: $\mathbb{U} := \mathbb{R}^d$, $\mathbb{C}(\mathbf{x}, \mathbf{x}, \mathbf{u}) := \mathbb{V}(\mathbf{x}) + \frac{1}{N} \sum W(\mathbf{x} - \mathbf{x}(\omega)) + \psi(\mathbf{u})$

We introduce the running and final cost functions

 $C = C(x, x, u) : \mathbb{R}^d \times (\mathbb{R}^d)^N \times \underline{\mathbb{U}} \to \mathbb{R}, \qquad C_T = C_T(x, x) : \mathbb{R}^d \times (\mathbb{R}^d)^N \to \mathbb{R}$

and we want to minimize the total cost

$$\mathcal{J}(\boldsymbol{X},\boldsymbol{U}) := \frac{1}{N} \sum_{\omega=1}^{N} \int_{0}^{T} \boldsymbol{C}(\boldsymbol{X}_{t}(\omega),\boldsymbol{X}_{t},\boldsymbol{U}_{t}(\omega)) \, dt + \frac{1}{N} \sum_{\omega=1}^{N} \boldsymbol{C}(\boldsymbol{X}_{T}(\omega),\boldsymbol{X}_{T}).$$

among all the admissible pairs $\,\,(X, {\color{black} U}) \in \mathcal{A}^{N}(X_{0})$ solving

$$\frac{d}{dt}X_t(\omega)=F(X_t(\omega),\boldsymbol{X}_t,\boldsymbol{U}_t(\omega)),\quad t\in(0,T),\quad X_{t=0}(\omega)=X_0(\omega).$$

$$\begin{split} & \text{Value function:} \qquad \mathcal{V}^{N}(\textbf{X}_{0}) := \inf \left\{ \mathcal{J}(\textbf{X},\textbf{U}) : (\textbf{X},\textbf{U}) \in \mathcal{A}^{N}(\textbf{X}_{0}) \right\} \\ & \text{mple:} \qquad \mathbb{U} := \mathbb{R}^{d}, \quad \textbf{C}(x,\textbf{x},\textbf{u}) := V(x) + \frac{1}{N}\sum_{\omega=1}^{N} \mathcal{W}(x - \textbf{x}(\omega)) + \psi(\textbf{u}) \end{split}$$

Example

Every **permutation** $\sigma \in Sym(N)$ of the set of indices $\Omega^N = \{1, 2, \cdots, N\}$ acts on a vector $\mathbf{x} \in (\mathbb{R}^d)^N$ just by

a permutation of its components: $\sigma_{\sharp} \mathbf{x} = (\mathbf{x}(\sigma(1)), \mathbf{x}(\sigma(2)), \cdots, \mathbf{x}(\sigma(N)))$

Invariance: $F(x, \sigma_{\sharp} x, u) = F(x, x, u), \quad C(x, \sigma_{\sharp} x, u) = C(x, x, u), \quad C_{T}(x, \sigma_{\sharp} x) = C_{T}(x, x).$

We may interpret a particle system as a dynamical system in the quotient space $(\mathbb{R}^d)^N/\sim$,

 $\mathbf{x} \sim \mathbf{y} \ \Leftrightarrow \ \mathbf{y} = \sigma_\sharp \mathbf{x} \quad \text{ for some } \sigma \in \mathsf{Sym}(\mathsf{N})$

Equivalently, we may associate to \mathbf{x} the discrete measure

$$\mu[x] := \frac{1}{N} \sum_{\omega=1}^{N} \delta_{x(\omega)} = x_\sharp \mathbb{P}^N \quad \mathbb{P}^N = \frac{1}{N} \sum_{\omega=1}^{N} \delta_{\omega} \text{ uniform discrete measure in } \Omega^N := \{1, \cdots, N\}.$$

 $F(x,x,u) \rightsquigarrow F(x,\mu[x],u), \quad C(x,x,u) \rightsquigarrow C(x,\mu[x],u), \qquad C_T(x,x,u) \rightsquigarrow C_T(x,\mu[x])$

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a permutation of its components: $\sigma_{\sharp} \mathbf{x} = (\mathbf{x}(\sigma(1)), \mathbf{x}(\sigma(2)), \cdots, \mathbf{x}(\sigma(N)))$

 $\label{eq:invariance:} \begin{array}{ll} \mathsf{F}(x,\sigma_\sharp x, u) = \mathsf{F}(x,x,u), \quad \mathsf{C}(x,\sigma_\sharp x, u) = \mathsf{C}(x,x,u), \quad \mathsf{C}_\mathsf{T}(x,\sigma_\sharp x) = \mathsf{C}_\mathsf{T}(x,x). \end{array}$

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Equivalently, we may associate to x the discrete measure

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 $F(x,x,u) \rightsquigarrow F(x,\mu[x],u), \quad C(x,x,u) \rightsquigarrow C(x,\mu[x],u), \qquad C_T(x,x,u) \rightsquigarrow C_T(x,\mu[x])$

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We may interpret a particle system as a dynamical system in the quotient space $(\mathbb{R}^d)^N / \sim$,

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Equivalently, we may associate to \mathbf{x} the discrete measure

$$\mu[\pmb{x}] := \frac{1}{N} \sum_{\omega=1}^N \delta_{\pmb{x}(\omega)} = \pmb{x}_{\sharp} \mathbb{P}^N \quad \mathbb{P}^N = \frac{1}{N} \sum_{\omega=1}^N \delta_\omega \text{ uniform discrete measure in } \Omega^N := \{1, \cdots, N\}.$$

 $\mathsf{F}(x, \boldsymbol{x}, \boldsymbol{u}) \rightsquigarrow \mathsf{F}(x, \boldsymbol{\mu}[\boldsymbol{x}], \boldsymbol{u}), \quad \mathbf{C}(x, \boldsymbol{x}, \boldsymbol{u}) \rightsquigarrow \mathbf{C}(x, \boldsymbol{\mu}[\boldsymbol{x}], \boldsymbol{u}), \qquad \mathbf{C}_\mathsf{T}(x, \boldsymbol{x}, \boldsymbol{u}) \rightsquigarrow \mathbf{C}_\mathsf{T}(x, \boldsymbol{\mu}[\boldsymbol{x}])$

From discrete to continuous

We suppose that we can extend by continuity the functions F, C, C_T to $\mathscr{P}_p(\mathbb{R}^d)$:

$$F(\mathbf{x}, \mathbf{x}, \mathbf{u}) = f(\mathbf{x}, \mathbf{u}) + \frac{1}{N} \sum_{\omega=1}^{N} g(\mathbf{x} - \mathbf{x}(\omega)) = f(\mathbf{x}, \mathbf{u}) + \int g(\mathbf{x} - \mathbf{y}) d\mu[\mathbf{x}](\mathbf{y})$$

$$\downarrow \downarrow$$

$$F(\mathbf{x}, \mu, \mathbf{u}) = f(\mathbf{x}, \mathbf{u}) + \int g(\mathbf{x} - \mathbf{y}) d\mu(\mathbf{y})$$

$$\mathbf{C}(\mathbf{x}, \mathbf{x}, \mathbf{u}) = \frac{1}{N} \sum_{\omega=1}^{N} W(\mathbf{x} - \mathbf{x}(\omega)) + \psi(\mathbf{u}) = \int W(\mathbf{x} - \mathbf{y}) d\mu[\mathbf{x}](\mathbf{y}) + \psi(\mathbf{u})$$

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$$\mathbf{C}(\mathbf{x}, \mu, \mathbf{u}) = \int W(\mathbf{x} - \mathbf{y}) d\mu(\mathbf{y}) + \psi(\mathbf{u})$$

 $\mu_n \to \mu \quad \Rightarrow \quad F(x, \mu_n, u) \to F(x, \mu, u), \quad C(x, \mu_n, u) \to C(x, \mu, u).$

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• Well posedness of the finite particle control problem.

- How to characterize and solve the limit problem: mean field optimal control and its Lagrangian, Eulerian, and Kantorovich formulations.
- Can we pass to the limit in the value function as $N
 ightarrow \infty$
- Can we pass to the limit in the **minimizers** as $N \to \infty$

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<u>MFOC:</u> Fornasier-Solombrino (Mean-field optimal control), Bongini-Fornasier, Fornasier-Piccoli-Rossi, Caponigro-Fornasier-Piccoli-Trelat, Albi-Choi-Fornasier, Fornasier-Piccoli-Rossi, ...

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$$D\big((x_1,\mu_1),(x_2,\mu_2)\big) := \Big(|x_1-x_2|^p + W_p^p(\mu_1,\mu_2)\Big)^{1/p}, \quad p \in [1,+\infty) \text{ fixed}.$$

- Continuity: F, C, C_T are continuous in $\mathbb{R}^d \times \mathscr{P}_p(\mathbb{R}^d) \times \mathbb{U}$ with p-growth.
- (Uniqueness) For every u ∈ U the map (x, μ) → F(x, μ, u), are D-Lipschitz with uniform Lipschitz constant (it would be possible to relax this condition assuming dissipativity)
- Coercivity: U is compact (or, in the additive case, $C(x, \mu, u) = C(x, \mu) + \psi(u), \psi$ superlinear with compact sublevels).

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Convexity: \mathbb{U} is convex, F is affine w.r.t. u, C is convex w.r.t. u.

Convex relaxation: $\mathbb{S} := (\mathbb{U}, \mathbb{F}, \mathbb{C}, \mathbb{C}_{\mathbb{T}}) \quad \rightsquigarrow \quad \mathbb{S}_{\mathbb{R}} := (\mathscr{U}, \mathscr{F}, \mathscr{C}, \mathscr{C}_{\mathbb{T}}):$

$$\begin{split} \mathscr{U} &:= \mathscr{P}(\mathbb{U}), \quad \mathscr{F}(\mathbf{x}, \mu, \sigma) := \int_{\mathbb{U}} \mathbb{P}(\mathbf{x}, \mu, u) \, d\sigma(u), \quad \mathscr{C}(\mathbf{x}, \mu, \sigma) := \int_{\mathbb{U}} \mathbb{C}(\mathbf{x}, \mu, u) \, d\sigma(u), \quad \mathscr{C}_{\mathsf{T}} := \mathbb{C}_{\mathsf{T}}. \\ \mathbf{S}_{\mathsf{R}} &:= (\mathscr{P}(\mathbb{U}), \mathscr{F}, \mathscr{C}, \mathscr{C}_{\mathsf{T}}) \text{ is a convex system extending S:} \\ \{\delta_{\mathfrak{u}} : \mathfrak{u} \in \mathbb{U}\} \subset \mathscr{U}, \end{split}$$

 $\mathscr{F}(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\delta}_{\mathbf{u}}) := F(\mathbf{x}, \boldsymbol{\mu}, \mathbf{u}), \quad \mathscr{C}(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\delta}_{\mathbf{u}}) := \mathbf{C}(\mathbf{x}, \boldsymbol{\mu}, \mathbf{u}).$

Convexity: \mathbb{U} is convex, F is affine w.r.t. u, C is convex w.r.t. u.

 $\textbf{Convex relaxation: } \mathbb{S} := (\textbf{U}, \textbf{F}, \textbf{C}, \textbf{C}_{T}) \quad \rightsquigarrow \quad \mathbb{S}_{R} := (\boldsymbol{\mathscr{U}}, \mathscr{F}, \mathscr{C}, \mathscr{C}_{T}) :$

$$\mathscr{U} := \mathscr{P}(\mathbb{U}), \quad \mathscr{F}(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\sigma}) := \int_{\mathbb{U}} F(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{u}) \, \mathrm{d}\boldsymbol{\sigma}(\boldsymbol{u}), \quad \mathscr{C}(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{\sigma}) := \int_{\mathbb{U}} \mathbf{C}(\mathbf{x}, \boldsymbol{\mu}, \boldsymbol{u}) \, \mathrm{d}\boldsymbol{\sigma}(\boldsymbol{u}), \quad \mathscr{C}_{\mathsf{T}} := \mathbf{C}_{\mathsf{T}}.$$

 $\mathbb{S}_{\mathbb{R}} := (\mathscr{P}(\mathbb{U}), \mathscr{F}, \mathscr{C}, \mathscr{C}_{\mathbb{T}})$ is a convex system extending S: $\{\delta_{\mathfrak{u}} : \mathfrak{u} \in \mathbb{U}\} \subset \mathscr{U},$

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Optimal control for finite particles: convex setting

Recall:

$$\mathcal{J}(\mathbf{X},\mathbf{U}) := \frac{1}{N} \sum_{\omega=1}^{N} \int_{0}^{T} \mathbf{C}(\mathbf{X}_{t}(\omega),\mathbf{X}_{t},\mathbf{U}_{t}(\omega)) \, dt + \frac{1}{N} \sum_{\omega=1}^{N} \mathbf{C}(\mathbf{X}_{T}(\omega),\mathbf{X}_{T}).$$

among all the admissible pairs $(\boldsymbol{X},\boldsymbol{U})\in \boldsymbol{\mathcal{A}}^N(\boldsymbol{X}_0)$ solving

$$\frac{d}{dt}X_t(\omega) = F(X_t(\omega), \mathbf{X}_t, \mathbf{U}_t(\omega)), \quad t \in (0, T), \quad X_{t=0}(\omega) = X_0(\omega).$$

Value function:

$$\mathcal{V}^{\mathsf{N}}(\boldsymbol{X}_{0}):= \inf \left\{ \mathcal{J}(\boldsymbol{X},\boldsymbol{U}): (\boldsymbol{X},\boldsymbol{U}) \in \mathcal{A}^{\mathsf{N}}(\boldsymbol{X}_{0}) \right\}$$

Theorem

In the **convex** setting the infimum is attained: there exists an optimal control U and a corresponding optimal trajectory X such that

 $(\mathbf{X}, \mathbf{U}) \in \mathcal{A}^{N}(X_{0}), \quad \mathcal{J}(\mathbf{X}, \mathbf{U}) = \mathcal{V}^{N}(X_{0})$

Chattering and convex relaxation

In the relaxed system $\mathbb{S}_R = (\mathcal{U}, \mathcal{F}, \mathcal{C}, \mathcal{C}_T)$ the controls $\sigma_t(\omega) = \sigma_{t,\omega}$ are probability measures in $\mathcal{U} = \mathscr{P}(\mathbb{U})$

$$\mathcal{J}_{\mathsf{R}}(\boldsymbol{X},\boldsymbol{\sigma}) := \frac{1}{N} \sum_{\omega=1}^{N} \int_{0}^{\mathsf{T}} \boxed{\int_{\boldsymbol{U}} \boldsymbol{C}(X_{\mathsf{t}}(\omega),\boldsymbol{X}_{\mathsf{t}},\boldsymbol{u}) \, d\boldsymbol{\sigma}_{\mathsf{t},\omega}(\boldsymbol{u})} \, d\mathsf{t} + \frac{1}{N} \sum_{\omega=1}^{N} \boldsymbol{C}(X_{\mathsf{T}}(\omega),\boldsymbol{X}_{\mathsf{T}}).$$

The admissible pairs $({f X},\sigma)\in {\cal A}_{\sf R}^{
m N}({f X}_0)$ solve

$$\frac{d}{dt}X_t(\omega) = \int_{U} F(X_t(\omega), X_t, u) \, d\sigma_{t, \omega}(u) \, , \quad t \in (0, T), \quad X_{t=0}(\omega) = X_0(\omega).$$

Value function:

$$V_R(X_0):= \inf \left\{ \mathfrak{J}_R(X,\sigma): (X,\sigma) \in \mathcal{A}_R^N(X_0) \right\}$$

Chattering and convex relaxation

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Value function:

$$V_{\mathrm{R}}(\mathrm{X}_{0}) := \inf \left\{ \mathcal{J}_{\mathrm{R}}(\mathrm{X}, \sigma) : (\mathrm{X}, \sigma) \in \mathcal{A}^{\mathrm{N}}_{\mathrm{R}}(\mathrm{X}_{0})
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Chattering and convex relaxation

In the relaxed system $\mathbb{S}_R = (\mathcal{U}, \mathcal{F}, \mathcal{C}, \mathcal{C}_T)$ the controls $\sigma_t(\omega) = \sigma_{t,\omega}$ are probability measures in $\mathcal{U} = \mathscr{P}(\mathbb{U})$

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Value function:

$$V_{\mathsf{R}}(X_{\mathsf{0}}) := \mathsf{inf}\left\{ \mathcal{J}_{\mathsf{R}}(\boldsymbol{X}, \boldsymbol{\sigma}) : (\boldsymbol{X}, \boldsymbol{\sigma}) \in \mathcal{A}_{\mathsf{R}}^{\mathsf{N}}(X_{\mathsf{0}}) \right\}$$

Theorem

The minimum of the relaxed problem coincides with the Lagrangian value function:

$$\mathcal{V}^{\mathsf{N}}(X_{0}) = \mathcal{V}^{\mathsf{N}}_{\mathsf{R}}(X_{0}) = \min\left\{\mathcal{J}_{\mathsf{R}}(\mathbf{X}, \boldsymbol{\sigma}) : (\mathbf{X}, \boldsymbol{\sigma}) \in \mathcal{A}^{\mathsf{N}}_{\mathsf{R}}(X_{0})\right\}$$

Mean field control: Lagrangian formulation

Particles are parametrized by an infinite probability parameter space $(\Omega, \mathfrak{B}, \mathbb{P})$:

$$\{1, 2, \cdots, N\} \rightsquigarrow \Omega, \qquad \frac{1}{N} \sum_{\omega} \delta_{\omega} \rightsquigarrow \mathbb{P}$$

 (Ω, \mathfrak{B}) is a standard Borel space (i.e. \mathfrak{B} is the Borel σ -algebra induced by a Polish topology on Ω) and \mathbb{P} is diffuse (every point has 0 mass). Canonical example: $\Omega = [0, 1], \mathbb{P} = \mathcal{L}^1$

Trajectories: $X \in L^p(\Omega; W^{1,p}(0, T; \mathbb{R}^d))$, $X_t \in L^p(\Omega; \mathbb{R}^d)$, $\mu_t = (X_t)_t \mathbb{P}$ is the law of X_t .

 $\frac{d}{dt}X_t(\omega) = F(X_t(\omega), \mu_t, U_t(\omega)) \quad \mu_t = (X_t)_\sharp \mathbb{P}, \quad X_{t=0} = X_0 \quad \text{in } L^p(\Omega; \mathbb{R}^d)$

 $U: (0,T) imes \Omega \to U$ is a $\mathcal{L}^1 imes \mathbb{P}$ -measurable map, $(X,U) \in \mathcal{A}_L(X_0)$.

 $\mathfrak{J}(X,U) := \int_\Omega \int_0^T C(X_t(\omega),\mu_t,U_t(\omega))\,dt\,d\mathbb{P}(\omega) + \int_\Omega C_T(X_T(\omega),\mu_T)\,d\mathbb{P}(\omega)$

 $\mathcal{V}_L(X_0):= \inf \left\{ \mathscr{J}(X,U): (X,U) \in \mathcal{A}_L(X_0) \right\}$

Mean field control: Lagrangian formulation

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 $\label{eq:constraint} \mbox{Trajectories: } \boldsymbol{X} \in L^p(\Omega; W^{1,p}(0,T;\mathbb{R}^d)), \ \boldsymbol{X}_t \in L^p(\Omega;\mathbb{R}^d), \ \boldsymbol{\mu}_t = (\boldsymbol{X}_t)_{\sharp}\mathbb{P} \ \mbox{is the law of } \boldsymbol{X}_t.$

 $\boxed{\frac{d}{dt}X_t(\omega)=F(X_t(\omega),\mu_t,\textbf{U}_t(\omega))} \quad \mu_t=(X_t)_{\sharp}\mathbb{P}, \quad \textbf{X}_{t=0}=X_0 \quad \text{in } L^p(\Omega;\mathbb{R}^d).$

 $\label{eq:constraint} \boldsymbol{\underline{U}}:(0,T)\times\Omega\rightarrow \boldsymbol{\underline{U}} \text{ is a } \mathcal{L}^1\times \mathbb{P}\text{-measurable map, } (\boldsymbol{X},\boldsymbol{\underline{U}})\in \mathcal{A}_L(X_0).$

 $\vartheta(\mathbf{X},\mathbf{U}) := \int_{\Omega} \int_{0}^{T} \mathbf{C}(\mathbf{X}_{t}(\omega),\mu_{t},\mathbf{U}_{t}(\omega)) \, dt \, d\mathbb{P}(\omega) + \int_{\Omega} \mathbf{C}_{\mathsf{T}}(\mathbf{X}_{\mathsf{T}}(\omega),\mu_{\mathsf{T}}) \, d\mathbb{P}(\omega)$

 $\mathbb{V}_{\mathbf{L}}(\mathrm{X}_{0}):=\inf\left\{ \mathcal{J}(\mathbf{X},\mathbf{U}):(\mathbf{X},\mathbf{U})\in\mathcal{A}_{\mathbf{L}}(\mathrm{X}_{0})
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Mean field control: Lagrangian formulation

Particles are parametrized by an infinite probability parameter space $(\Omega, \mathfrak{B}, \mathbb{P})$:

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$$\frac{d}{dt}X_t(\omega) = F(X_t(\omega), \mu_t, \boldsymbol{U}_t(\omega)) \qquad \mu_t = (X_t)_{\sharp} \mathbb{P}, \quad \boldsymbol{X}_{t=0} = X_0 \quad \text{in } L^p(\Omega; \mathbb{R}^d).$$

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$$\mathcal{J}(\boldsymbol{X},\boldsymbol{U}) := \int_{\Omega} \int_{0}^{T} \boldsymbol{C}(\boldsymbol{X}_{t}(\boldsymbol{\omega}),\boldsymbol{\mu}_{t},\boldsymbol{U}_{t}(\boldsymbol{\omega})) \, dt \, d\mathbb{P}(\boldsymbol{\omega}) + \int_{\Omega} \boldsymbol{C}_{T}(\boldsymbol{X}_{T}(\boldsymbol{\omega}),\boldsymbol{\mu}_{T}) \, d\mathbb{P}(\boldsymbol{\omega})$$

$$\mathcal{V}_{\mathbf{L}}(\mathbf{X}_{0}) := \inf \left\{ \mathcal{J}(\mathbf{X}, \mathbf{U}) : (\mathbf{X}, \mathbf{U}) \in \mathcal{A}_{\mathbf{L}}(\mathbf{X}_{0}) \right\}$$

N-particle system:

 $\mathfrak{B}^N \subset \mathfrak{B}$ is a finite algebra generated by a **finite partition** $\mathfrak{B}^N = (B^1, \cdots, B^N)$ of Ω with $\mathbb{P}(B^i) = 1/N$.

 \bm{X}_t and \bm{U}_t are \mathfrak{B}^N -measurable, i.e. constant on each set B^i of the partition.

 $X_t(\omega)$, $U_t(\omega)$ as $\omega \in B^i$, represents the i-th particle driven by its control.

Starting from $\mathbb{S} = (\mathbb{U}, F, C, C_T)$ construct $\mathbb{S}_R = (\mathcal{U}, \mathcal{F}, \mathcal{C}, \mathcal{C}_T)$ and the corresponding $\mathcal{A}_{RL}(X_0)$, $\mathcal{J}_{RL}(X, \sigma)$ associated to the measure-valued controls $\sigma : (0, T) \times \Omega \to \mathscr{P}(\mathbb{U})$.

 $\mathbb{S}_{R} = (\mathcal{U}, \mathcal{F}, \mathcal{C}, \mathcal{C}_{T})$ is a **convex** mean field optimal control problem.

Theorem (CLOS)

For every $X_0 \in L^p(\Omega; \mathbb{R}^d)$

$$V_{\mathbf{L}}(X_0) = V_{\mathbf{RL}}(X_0)$$

Starting from $\mathbb{S} = (\mathbb{U}, F, C, C_T)$ construct $\mathbb{S}_R = (\mathscr{U}, \mathscr{F}, \mathscr{C}, \mathscr{C}_T)$ and the corresponding $\mathcal{A}_{RL}(X_0)$, $\mathcal{J}_{RL}(\mathbf{X}, \boldsymbol{\sigma})$ associated to the measure-valued controls $\boldsymbol{\sigma} : (0, T) \times \Omega \to \mathscr{P}(\mathbb{U})$.

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Theorem (CLOS)

For every $X_0 \in L^p(\Omega; \mathbb{R}^d)$

$$V_{\boldsymbol{L}}(X_0) = V_{\boldsymbol{R}\boldsymbol{L}}(X_0)$$

Remark

There are cases when the infimum is not attained, even if the problem is convex!!

Lack of compactness w.r.t. $\omega \in \Omega$ of trajectories and controls.

A very simple setting: $X_0 \in L^2(\Omega; \mathbb{R}^d)$ with law $\mu_0 = (X_0)_{\sharp} \mathbb{P}$ and μ_1 in $\mathscr{P}_2(\mathbb{R}^d)$ are given.

$$\mathsf{T} = \mathsf{1}, \quad \mathbb{U} = \mathbb{R}^{d}, \quad \mathsf{F}(\mathsf{x}, \boldsymbol{\mu}, \boldsymbol{u}) = \boldsymbol{u}, \quad C(\mathsf{x}, \boldsymbol{\mu}, \boldsymbol{u}) := |\boldsymbol{u}|^{2}, \quad C_{\mathsf{T}}(\mathsf{x}, \boldsymbol{\mu}) := \begin{cases} 0 & \text{if } \boldsymbol{\mu} = \boldsymbol{\mu}_{\mathsf{1}}, \\ +\infty & \text{otherwise.} \end{cases}$$

ODE System: $\dot{X}_t(\omega) = U_t(\omega)$, the trajectories minimizing $\int_0 |U_t(\omega)|^2 dt = \int_0 |\dot{X}_t(\omega)|^2 dt$, are segments joining $X_0(\omega)$ to $X_1(\omega)$.

$$\mathcal{V}_L(X_0) = \inf \left\{ \int |X_1(\omega) - X_0(\omega)|^2 \, d\mathbb{P}(\omega) : (X_1)_\sharp \mathbb{P} = \mu_1 \right\} = W_2^2(\mu_0, \mu_1)$$

If μ_0 is diffuse but **concentrated on a set of codimension** ≥ 1 **in** \mathbb{R}^d there are **examples where** the infimum is not attained. Take $\Omega := \mathbb{R}^d$, $\mathbb{P} = \mu_0$, $X_0(\omega) = \omega$

$$\mathcal{V}_{\mathsf{L}}(X_0) = \left[\inf\left\{\int |X(\omega) - \omega|^2 \, \mathsf{d}\mu_0(\omega) : (X)_\sharp \mu_0 = \mu_1
ight\} \rightsquigarrow \mathsf{Monge formulation}
ight.$$

More regular $C_{\mathsf{T}}\colon C_{\mathsf{t}}(x,\mu):=W_2^2(\mu,\mu_1)$, Monge formulation of the barycenter problem

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$$\mathbb{V}_{L}(X_{0}) = \inf \left\{ \bigcup |X_{1}(\omega) - X_{0}(\omega)|^{2} \, d\mathbb{P}(\omega) : (X_{1})_{\sharp}\mathbb{P} = \mu_{1} \right\} = W_{2}^{2}(\mu_{0}, \mu_{1})$$

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If μ_0 is diffuse but **concentrated on a set of codimension** $\geqslant 1$ in \mathbb{R}^d there are **examples where** the infimum is not attained. Take $\Omega := \mathbb{R}^d$, $\mathbb{P} = \mu_0$, $X_0(\omega) = \omega$

$$\mathcal{V}_{\mathbf{L}}(X_0) = \left. \inf \left\{ \int |X(\omega) - \omega|^2 \, d\mu_0(\omega) : (X)_\sharp \mu_0 = \mu_1 \right\} \rightsquigarrow$$
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If μ_0 is diffuse but concentrated on a set of codimension ≥ 1 in \mathbb{R}^d there are examples where the infimum is not attained. Take $\Omega := \mathbb{R}^d$, $\mathbb{P} = \mu_0$, $X_0(\omega) = \omega$

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More regular C_T : $C_t(x, \mu) := W_2^2(\mu, \mu_1)$, Monge formulation of the barycenter problem.

Eulerian (a.k.a. Benamou-Brenier) formulation: look to the evolution of $\mu_t = (X_t)_{\sharp} \mathbb{P}$ driven by a control field $\mathbf{u} : (0, T) \times \mathbb{R}^d \to \mathbb{R}^d$. $\mathcal{A}_E(\mu_0)$ is the set of pairs (μ, \mathbf{u}) solving the continuity equation

$$\partial_t \mu_t + \nabla \cdot (\boldsymbol{\mathfrak{u}}_t \mu_t) = 0 \quad \text{ in } (0,T) \times \mathbb{R}^d$$

minimizing the cost

$$\mathcal{J}_{\mathsf{E}}(\boldsymbol{\mu},\boldsymbol{u}) := \int_{0}^{T} \int |\boldsymbol{u}_{t}|^{2} \, d\boldsymbol{\mu}_{t} \, dt$$

Kantorovich formulation: look to the couplings $\mathbb{G} = (X_0, X_1)_{\sharp}\mathbb{P}$ of μ_0, μ_1 in $\mathscr{P}_2(\mathbb{R}^d \times \mathbb{R}^d)$ and minimize the cost

$$\begin{split} \mathcal{J}_{\mathbf{K}}(\mathbb{G}) &:= \int |\mathbf{x}_1 - \mathbf{x}_0|^2 \, d\mathbf{G}(\mathbf{x}_0, \mathbf{x}_1) \\ (\mu_0) &= \min_{\mathcal{A}_{\mathbf{K}}(\mu_0)} \mathcal{J}_{\mathbf{E}} = \mathcal{V}_{\mathbf{K}}(\mu_0) = \min_{\mathcal{A}_{\mathbf{K}}(\mu_0)} \mathcal{J}_{\mathbf{K}} = \mathcal{V}_{\mathbf{L}}(\mathbf{x}_0, \mathbf{x}_0) \end{split}$$

Eulerian (a.k.a. Benamou-Brenier) formulation: look to the evolution of $\mu_t = (X_t)_{\sharp} \mathbb{P}$ driven by a control field $\mathbf{u} : (0, T) \times \mathbb{R}^d \to \mathbb{R}^d$. $\mathcal{A}_E(\mu_0)$ is the set of pairs (μ, \mathbf{u}) solving the continuity equation

$$\left| \vartheta_t \mu_t + \nabla \cdot (\boldsymbol{u}_t \mu_t) = 0 \right| \quad \text{in } (0,T) \times \mathbb{R}^d$$

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Variational Г convergence

Suppose that $X_0^N \to X_0^\infty$ in $L^p(\Omega; \mathbb{R}^d)$, X_0^N are initial parametrizations of a N-particle system. $(\mathbf{X}^N, \mathbf{U}^N) \in \mathcal{A}_L(X_0^N)$ are minimizers, \mathfrak{B}^N -measurable. In order to pass to the limit in V_L we need

- Compactness for trajectories-controls w.r.t. some notion of convergence: there exists a subsequence $(\mathbf{X}^{N(k)}, \mathbf{U}^{N(k)})$ converging to some limit $(\mathbf{X}^{\infty}, \mathbf{U}^{\infty}) \in \mathcal{A}_{L}(X_{0}^{\infty})$.
- Γ -lim inf **inequality:** along every sequence (X^{κ}, U^{κ}) converging to (X^{∞}, U^{∞}) lim inf_{k→∞} $\mathcal{J}_{L}(X^{N(k)}, U^{N(k)}) \ge \mathcal{J}_{L}(X^{\infty}, U^{\infty})$
- Γ -lim sup **estimate:** if $(X^{\infty}, \mathbb{U}^{\infty}) \in \mathcal{A}_{L}(X_{0}^{\infty})$ then there exists $(X^{N}, \mathbb{U}^{N}) \in \mathcal{A}_{L}(X_{0}^{N})$ such that $\limsup_{N \to \infty} \mathcal{J}_{L}(X^{N}, \mathbb{U}^{N}) \leqslant \mathcal{J}_{L}(X^{\infty}, \mathbb{U}^{\infty})$.



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$$\mu_0 = (X_0)_{\sharp} \mathbb{P}, \quad \mu_t = (X_t)_{\sharp} \mathbb{P} \quad \text{in } \mathscr{P}_p(\mathbb{R}^d)$$

The control is a Borel map $u:(0,T)\times \mathbb{R}^d \to \underline{U}.$

When (X_t, \boldsymbol{U}_t) are piecewise constant on the partition \mathcal{B}^N the measure μ_t is discrete: $\mu_t = \frac{1}{N} \sum_{i=1}^N \delta_{X_t(B^i)}$



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 μ is a curve in $AC^p(0,T; \mathscr{P}_p(\mathbb{R}^d))$ solving the continuity equation

$$\partial_t \mu_t + \nabla \cdot (\nu_t \mu_t) = 0 \quad \text{in } (0,T) \times \mathbb{R}^d, \quad \nu_t(x) = F(x,\mu_t,u_t(x)) \ \ \ \ \mu_{t=0} = \mu_0$$

 $(\boldsymbol{\mu}, \boldsymbol{u}) \rightsquigarrow \mathcal{A}_{E}(\boldsymbol{\mu}_{0}).$

$$\begin{split} \mathfrak{J}_{\mathsf{E}}(\mu,\mathfrak{u}) &:= \int_0^T \int_{\mathbb{R}^d} C(x,\mu_t,\mathfrak{u}_t(x)) \, d\mu_t(x) \, dt + \int_{\mathbb{R}^d} C_T(x,\mu_T) \, d\mu_T(x) \\ & V_{\mathsf{E}}(\mu_0) := \inf \Big\{ \mathfrak{J}_{\mathsf{E}}(\mu,\mathfrak{u}) : (\mu,\mathfrak{u}) \in \mathcal{A}_{\mathsf{E}}(\mu_0) \Big\}. \end{split}$$

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Compactness in the Eulerian problem

- Given a sequence $(\mu^k, \mathbf{u}^k) \in \mathcal{A}_E(\mu_0^k)$ such that $\mu_0^k \to \mu_0$, we pass to the relaxed formulation via the **Young measure** $\sigma^k := (id \times \mathbf{u}^k)_{\sharp} \mu^k$. Continuity equation:
 - $\iiint \left(\partial_t \zeta(t,x) + \nabla \zeta(t,x) \cdot F(x,\mu_t,\textbf{u}) \right) d\sigma^k(t,x,\textbf{u}) = 0 \quad \text{for every } \zeta \in C^1_c((0,T) \times \mathbb{R}^d).$

$$\mathcal{J}_{\mathsf{E}}(\mu^{k}, \mathbf{u}^{k}) = \iiint \mathbf{C}(x, \mu_{t}, \mathbf{u}) d\mathbf{\sigma}^{k}(t, x, \mathbf{u}) + \int \mathbf{C}_{\mathsf{T}}(x, \mu_{t}) d\mu_{\mathsf{T}}(x) d\mathbf{u}_{\mathsf{T}}(x) d\mathbf{u}_{\mathsf{$$

- Young measure solutions (including controls) are tight: the marginal w.r.t. to t is the Lebesgue measure on (0, T), the marginal w.r.t. u is supported in the compact set U, the marginal w.r.t. x are controlled by momentum estimates. Uniform equicontinuity in time for $\mu_{\rm t}^{\rm k}$ in $\mathscr{P}_{\rm p}(\mathbb{R}^{\rm d})$.
- From a limit measure σ^{∞} w.r.t. the topology of $\mathscr{P}((0,T) \times \mathbb{R}^d \times \mathbb{U})$ we recover μ^{∞} as the marginal of σ w.r.t. (t, x)

 u^∞ as the barycenter w.r.t. u, i.e. by disintegrating $\sigma^\infty = \int \sigma_{t,x} \, d\mu^\infty$ and

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The limit continuity equation:

$$\iiint \left(\vartheta_t \zeta(t,x) + \nabla \zeta(t,x) \cdot F(x,\mu_t,\textbf{u}) \right) d \textbf{\sigma}^\infty(t,x,\textbf{u}) = 0 \quad \text{for every } \zeta \in C^1_c((0,T) \times \mathbb{R}^d).$$

yields

$$\partial_{\mathbf{t}} \boldsymbol{\mu}^{\infty} + \nabla \cdot (\boldsymbol{u}^{\infty} \, \boldsymbol{\mu}^{\infty}) = \boldsymbol{0}, \quad (\boldsymbol{\mu}^{\infty}, \boldsymbol{u}^{\infty}) \in \mathcal{A}_{\mathsf{E}}(\boldsymbol{\mu}_{0})$$

. Convexity and Jensen inequality yield

$$\liminf_{k \to \infty} \mathcal{J}_{\mathsf{E}}(\mu^k, \mathbf{u}^k) \geqslant \mathcal{J}_{\mathsf{E}}(\mu, \mathbf{u})$$

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Theorem

In the convex setting there exists a minimizer of the Eulerian MFOC

$$V_{\mathsf{E}}(\mu_0) = \min \left\{ \mathcal{J}_{\mathsf{E}}(\mu, \mathbf{u}) : (\mu, \mathbf{u}) \in \mathcal{A}_{\mathsf{E}}(\mu_0) \right\}$$

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and

 $\liminf_{N\to\infty} V_E(\mu_0^N) \geqslant V_E(\mu_0).$

Moreover, if $X_0^N \to X_0$ strongly in $L^p(\Omega; \mathbb{R}^d)$ then $\mu_0^N \to \mu_0$ in $\mathscr{P}_p(\mathbb{R}^d)$ and $\liminf_{N \to \infty} V_L(X_0^N) \ge V_E(\mu_0).$

Lagrangian $^{\infty}$ vs Eulerian $^{\infty}$ 1

Theorem

Let $X_0 \in L^p(\Omega; \mathbb{R}^d)$ be given with $\mu_0 = (X_0)_{\sharp} \mathbb{P}$ and let $(\mathbf{X}, \mathbf{U}) \in \mathcal{A}_L(X_0)$. if $\boldsymbol{\sigma} := (\mathbf{X}, \mathbf{U})_{\sharp} (\mathcal{L}^1 \times \mathbb{P})$ and $\mu = \mathbf{X}_{\sharp} (\mathcal{L}^1 \times \mathbb{P})$, the barycenter \mathbf{u} of $\boldsymbol{\sigma}$ w.r.t. its marginal μ satisfies

 $(\mu, \boldsymbol{u}) \in \boldsymbol{\mathcal{A}}_E(\mu_0), \qquad J_L(\boldsymbol{X}, \boldsymbol{U}) \geqslant J_E(\mu, \boldsymbol{u})$

In particular

$$\mathcal{V}_L(\mathbb{X}_0) = \mathcal{V}_E(\mu_0)$$



$\Gamma\text{-lim}\ \text{sup}\ \text{inequality: the Kantorovich formulation}$

It is not clear how to prove directly the Γ -lim sup **inequality** at the Eulerian level. Moreover, it would be interesting to improve the inqualities between the Lagrangian and the Eulerian formulations.

As in Optimal Transport, it is useful to introduce a new formulation:



Kantorovich formulation

Path space: $\Gamma := C([0, T]; \mathbb{R}^d)$, curves in Γ are denoted by ω .

 X_t is the evaluation map at time $t : X_t(\boldsymbol{\omega}) := \boldsymbol{\omega}(t)$.

Controls depend on time t and on curves $\boldsymbol{\omega}$: $\boldsymbol{U}(t, \boldsymbol{\omega})$.

Given a probability measure \mathbb{G} on Γ , its time marginals $\mu_t = (X_t)_{\sharp}\mathbb{G}$ are evolving probability laws.

 $\mathcal{A}_{\mathbf{K}}(\mu_0)$ is given by pairs (G, U) such that G-almost every curve ω solves

$$\frac{d}{dt}\omega(t)=F(\omega(t),\mu_t,U_t(\omega)),\quad (X_0)_\sharp\mathbb{G}=\mu_0$$

The corresponding cost can be computed as

$$\mathfrak{J}_K(\mathbb{G}, U) = \int_\Gamma \int_0^T C(\boldsymbol{\omega}(t), \mu_t, U_t(\boldsymbol{\omega})) \, dt \, d\mathbb{G}(\boldsymbol{\omega}) + \int_{\mathbb{R}^d} C_T(x, \mu_T) \, d\mu_T(x) \quad \mu_t = (X_t)_t \mathbb{Q}_t$$

 $\mathbb{V}_{\mathbf{K}}(\mu_0) = \inf \Big\{ \mathcal{J}_{\mathbf{K}}(\mathbb{G}, \mathbb{U}) : (\mathbb{G}, \mathbb{U}) \in \mathcal{A}_{\mathbf{K}}(\mu_0) \Big\}.$

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Once G is fixed, the Kantorovich formulation yields a a particular version of a Lagrangian problem with

$$\Omega:=\Gamma,\quad \mathfrak{B}=\Big\{\text{Borel }\sigma\text{-algebra of }\Gamma\Big\},\quad \mathbb{P}=\mathbb{G}$$

However, in this formulation, we fix the maps X_t (the evaluation maps) and we allow for modifications of the reference measure $\mathbb P$.

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Eulerian vs Kantorovich (1)

Theorem

For every μ_0 and $(\mu, \mathbf{u}) \in \mathcal{A}_E(\mu_0)$ there exists $(\mathbb{G}, \mathbf{U}) \in \mathcal{A}_K(\mu_0)$ such that

 $\mu_{t} = (X_{t})_{\sharp}\mathbb{G}, \quad \mathcal{J}_{K}(\mathbb{G}, \mathbf{U}) = \mathcal{J}_{E}(\mu, \mathbf{u}).$

In particular $\mathcal{V}_{\mathbf{K}}(\mu_0) \leqslant \mathcal{V}_{\mathbf{E}}(\mu_0)$.

It is a consequence of the **superposition theorem**: given μ solving

$$\partial_t \mu +
abla \cdot (\mathbf{v}\mu) = 0, \quad \int_0^T \int |\mathbf{v}|^p \, d\mu_t \, dt < +\infty,$$

there exists a probability measure $\mathbb G$ on Γ such that

T is concentrated on solutions of $\omega'(t) = v_t(\omega_t)$, and $(X_t)_t \mathbb{G} = \mu_t$ in [0, T].

Since $v_t(\omega(t)) = F(\omega(t), \mu_t, u_t(\omega(t)))$ we can define $U_t(\omega) := u_t(\omega(t))$ and we obtain a pair $(\mathbb{G}, U) \in \mathcal{A}_K(\mu_0)$ with $\mathcal{J}_K(\mathbb{G}, U) = \mathcal{J}_E(\mu, u)$.

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Eulerian vs Kantorovich (2)

Since every solution $(\mathbb{G}, U) \in \mathcal{A}_{K}(\mu_{0})$ is a Lagrangian solution, we know that $V_{K}(\mu_{0}) \ge V_{E}(\mu_{0})$.



Every Lagrangian pair $(\mathbf{X}, \mathbf{U}) \in \mathcal{A}_{L}(X_{0})$ induces a Kantorovich pair with lower cost. Conversely, every Kantorovich pair can be approximated by Lagrangian pairs.

Theorem

For every $X_0 \in L^p(\Omega; \mathbb{R}^d)$ with $\mu_0 = (X_0)_{\sharp}\mathbb{P}$ and every $(\mathbb{G}, \mathbf{U}) \in \mathcal{A}_{\mathbf{K}}(\mu_0)$ there exists $(\mathbf{X}^n, \mathbf{U}^n) \in \mathcal{A}_{\mathbf{L}}(X_0)$ such that $\mathcal{J}_{\mathbf{L}}(\mathbf{X}^n, \mathbf{U}^n) \to \mathcal{J}_{\mathbf{K}}(\mathbb{G}, \mathbf{U})$. In particular $\mathcal{V}_{\mathbf{L}}(X_0) = \mathcal{V}_{\mathbf{K}}(\mu_0) = \mathcal{V}_{\mathbf{E}}(\mu_0)$.



Finite particle approximations

Theorem

Suppose that X_0^N is a family of N particles such that $\mu_0^N = (X_0^N)_{\sharp}\mathbb{P}$ converges to μ_0^∞ in $\mathscr{P}_p(\mathbb{R}^d)$ and let X_0^∞ such that $(X_0^\infty)_{\sharp}\mathbb{P} = \mu_0$. Then

$$\lim_{N\to\infty}\mathcal{V}_{L}(X_{0}^{N})=\lim_{N\to\infty}\mathcal{V}_{E}(\mu_{0}^{N})=\lim_{N\to\infty}\mathcal{V}_{K}(\mu_{0}^{N})=\mathcal{V}_{L}(X_{0}^{\infty})=\mathcal{V}_{E}(\mu_{0}^{\infty})=\mathcal{V}_{K}(\mu_{0}^{\infty})$$

