# Kantorovich operators and their ergodic properties

#### Nassif Ghoussoub, UBC (based on joint work with Malcolm Bowles)

Kantorovich Initiative Seminar, March 2023

- 1. Non-linear Kantorovich operators in analysis.
- 2. 1-homogenous Kantorovich operators and "zero cost" balayage.
- 3. Kantorovich operators and Choquet functional capacities.
- 4. Duality: Kantorovich operators, linear transfers, optimal balayage.
- 5. Weak KAM solutions/operators associated to Kantorovich operators.
- 6. Deterministic and stochastic Fathi-Mather theory.
- 7. Deterministic and stochastic Ergodic optimization.

#### Formal definition of Kantorovich operators

A Markov operator is a map  $T : C(Y) \to C(X)$  which is:

- 1. **positive**: if  $g \ge 0$  then  $Tg \ge 0$ .
- 2. linear:  $T(\lambda g_1 + \mu g_2) = \lambda T g_1 + \mu T^- g_2$ .
- 3. Markovian: T1 = 1.
- 4. **continuous**:  $g_n \to g$  in C(Y), then  $\lim_{n\to\infty} Tg_n = Tg$ .

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A backward Kantorovich operator  $T^- : C(Y) \rightarrow USC(X)$  which is

- 1. monotone increasing, i.e., if  $g_1 \leq g_2$ , then  $T^-g_1 \leq T^-g_2$ .
- 2. affine on the constants, i.e., for any  $c \in \mathbb{R}$  and  $g \in C(Y)$  $T^{-}(g+c) = T^{-}g + c$ .
- 3. convex, i.e., For any  $\lambda \in [0,1]$ , we have

$$T^-(\lambda g_1+(1-\lambda)g_2)\leq \lambda T^-g_1+(1-\lambda)T^-g_2$$

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4. lower semi-continuous, i.e., If  $g_n \to g$ , then  $\liminf_{n \to \infty} T^-g_n \ge T^-g$ .

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A forward Kantorovich operator is a map  $T^+$ :  $C(X) \rightarrow LSC(Y)$  that satisfies 1), 2), 3') (concave) and 4') (upper-semi-continuous).

1.  $U^-g = g \lor Tg$ , where T is Markov. Iterates lead to  $U^-_{\infty}g = \hat{g}$  the least T-superharmonic function above g.

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- 2. Plurisuperharmonic envelopes

$$U^{-}g(x) := \sup_{v \in \mathbb{R}^n} \bigg\{ \int_0^{2\pi} g(x + e^{i\theta}v) \frac{d\theta}{2\pi}; x + \bar{\Delta}v \subset O \bigg\},$$

Iterates lead to

$$U_{\infty}^{-}g(x) = \sup\left\{\int_{0}^{2\pi} g(P(e^{i\theta}))\frac{d\theta}{2\pi}; \ P \text{ polynomial}, \ P(\bar{\Delta}) \subset U, \ P(0) = x\right\}$$

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Image: Second second

3. Superharmonic envelope and Optimal stopping:

$$\begin{split} U^{-}g(x) &= \sup_{r \geq 0} \bigg\{ \int_{B} g(x+ry) \, dm(y); \, x+r\overline{B} \} \subset O \bigg\}, \\ U^{-}_{\infty}g(x) &:= \sup \Big\{ \mathbb{E}^{x} \Big[ g(B_{\tau}) \Big]; \, \tau \geq 0 \text{ stopping time}, \mathbb{E}^{x}[\tau] < +\infty \Big\}. \end{split}$$
  
The expectation  $\mathbb{E}^{x}$  refers to Brownian motions  $(B_{t})_{t}$  starting at  $x$ .

4. The filling scheme for a Markov operator T,  $U^-g = Tg^+ - g^-$ .

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4. The *filling scheme* for a Markov operator T,  $U^-g = Tg^+ - g^-$ .

All these are positively 1-homogenous Kantorovich operators,

$$U(\lambda f) = \lambda U(f)$$
 for all  $\lambda \ge 0$ .

### 1-homogenous Kantorovich operators

A typical 1-homogenous operator is

$$U^-g(x):=\sup\{\int_X g\,d\sigma;(x,\sigma)\in\mathcal{S}\},$$

where g is a reward function, and  $S \subset X \times \mathcal{P}(Y)$  is a "gambling house". • For each x,  $S_x = \{\sigma \in \mathcal{P}(X); (x, \sigma) \in S\} \neq \emptyset$  is the collection of distributions of gains available to a gambler having wealth x (Dubins-Savage, Dellacherie-Meyer).

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• When X = Y, a fair gambling houser is

$$U^{-}g(x) := \sup\{\int_{X} g \ d\sigma; \ \delta_{x} \prec_{\mathcal{A}} \sigma\},\$$

where A is the cone of convex l.s.c. functions on X (convex compact):

 $\mu \prec_{\mathcal{A}} \nu$  iff  $\int_{X} \phi d\mu \leq \int \phi d\nu$  for all  $\phi \in \mathcal{A}$ .

Convex order, Fair game, martingale, etc... A very particular case, but somewhat characterizes 1-homogenous Kantorovich operators from Y to X.

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# A representation of 1-homogenous Kantorovich operators

**Theorem:** The following are equivalent:

- 1. T is a 1-homogenous Kantorovich operator from C(Y) to USC(X).
- There exists a closed convex, stable under finite max, balayage cone A on the disjoint union X ∐ Y such that

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where for  $(\mu, \nu) \in \mathcal{P}(X) imes \mathcal{P}(Y)$ ,

 $\mu \prec_{\mathcal{A}} \nu \quad \text{iff} \quad \int_{X} \phi_{X} d\mu \leq \int_{Y} \phi_{Y} d\nu \text{ for all } \phi \in \mathcal{A}. \quad (\text{Restricted balayage}).$ 

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• The gambling house is then,

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where  $U_{\#}^{-}\mu(g) = \int_{X} U^{-}g \, d\mu$  for every  $g \in C(Y)$ . • If X = Y, then iterating  $Tu = u \vee U^{-}u$ , leads to an idempotent backward Kantorovich operator  $U_{\infty}^{-}$  (i.e.,  $U_{\infty}^{-} \circ U_{\infty}^{-} = U_{\infty}^{-}$ ) and a balayage cone  $\mathcal{A} \subset LSC(X)$  such that

$$U^-_{\infty}g(x) := \inf\{\phi(x); \phi \in -\mathcal{A}, \phi \geq g \text{ on } X\}.$$

# Non-homogenous Kantorovich operators

1. Ergodic optimization of symbolic dynamics:

$$U^{-}g(x) := g \circ \sigma(x) - A(x),$$

A is a given potential and  $\sigma$  is a point transformation. Its iterates lead to minimizing the action  $\mu \mapsto \int_{\mathbf{x}} A d\mu$  among all  $\sigma$ -invariant measures  $\mu$ .

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2. Optimal mass transport with cost function c(x, y): on  $X \times Y$ .

$$U^{-}g(x) = \sup_{y \in Y} \{g(y) - c(x, y)\} \text{ resp., } U^{+}f(y) = \inf_{x \in X} \{f(x) + c(x, y)\},$$

is then a backward (resp., forward Kantorovich operator): (Brenier transport):  $U^-g = -g^*$  resp.,  $U^+f = (-f)^*$ ,  $\phi^*$  is the Legendre transform.

3. Entropic regularization and Sinkhorn

$$T_{\nu}^{-}g(x) = \epsilon \log \int_{Y} e^{\frac{g(y) - c(x,y)}{\epsilon}} d\nu(y),$$

where  $\nu \in \mathcal{P}(X)$ , as well as the composition  $T_{\nu}^{-} \circ T_{\mu}^{-}$ , where  $\mu$  is another probability in  $\mathcal{P}(X)$ , whose iterates are the building clocks of the Sinkhorn algorithm.

If L is a Tonelli Lagrangian on TM

$$U^{-}g(x):=\sup\Big\{g(\gamma(1))-\int_{0}^{1}L(\gamma(s),\dot{\gamma}(s))\,ds;\gamma\in C^{1}([0,1),M);\gamma(0)=x\Big\},$$

To a state g at time 1, it associates the initial state of the viscosity solution for the associated backward Hamilton-Jacobi equation,

$$\begin{cases} \partial_t V + H(x, \nabla_x V) = 0 \text{ on } (0, 1) \times M \\ V(1, x) = g(x) \end{cases}$$

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All the above have corresponding forward Kantorovich operators.

#### One-sided Kantorovich operators

1. Stochastic control: One-sided Kantorovich operators appear in stochastic mass transfer problems. For example,

$$T^{-}g(x) = \sup_{X \in \mathcal{A}_{[0,1]}} \left\{ \mathbb{E}\left[g(X(1)) - \int_0^1 L(X(s), \beta_X(s, X)) ds \mid X(0) = x\right] \right\},$$

where  $\mathcal{A}_{[0,1]} := \{X : \Omega \to U; dX_t = \beta(t, X) dt + dW_t \text{ on } [0,1]\}$ .  $W_t$  is Weiner measure and the minimization is taken over all drifts  $\beta$ .

Under some assumptions on the Lagrangian L,

 $T^-g = J_g(0,x)$  where  $J_g$  is the initial state of the backward second order Hamilton-Jacobi equation

$$\begin{cases} \partial_t J(t,x) + \frac{1}{2}\Delta J(t,x) + H(x,\nabla J(t,x)) &= 0 \text{ in } (0,1) \times \mathbb{R}^d, \\ J(1,x) &= g(x) \text{ on } \mathbb{R}^d. \end{cases}$$

Where H is the Hamiltonian associated to L.

# Non-linear probability and potential theories?

1. A "non-linear potential theory" ?: A cost  $c : O \times O \to \mathbb{R} \cup \{+\infty\}$  is assigned to moving energy on a convex bounded domain in  $\mathbb{R}^d$ . The operator  $U^-g(x) = u_{g,x}$ , where  $u_{g,x}$  is the unique minimiser of

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2. Optimal stopping with cost: The operator  $U^-g = J_g(0, \cdot)$ , defined via the dynamic programming principle

$$U^{-}g(x) := \sup_{\tau \in \mathcal{R}^{x}} \Big\{ \mathbb{E}^{x} \Big[ g(B_{\tau}) - \int_{0}^{\tau} L(s, B_{s}) ds \Big] \Big\},$$

If  $t \to L(t, x)$  is decreasing, then  $U^-$  is an idempotent Kantorovich operator (i.e.,  $U^2 = U$ ).

 $U^-g(x)$  is actually a "variational solution" at time 0, for the quasi-variational Hamilton-Jacobi-Bellman equation:

$$\min\left\{\begin{array}{c}J(t,x)-g(x)\\-\frac{\partial}{\partial t}J(t,x)-\frac{1}{2}\Delta J(t,x)+L(t,x)\end{array}\right\}=0.$$

**Theorem:** The following are equivalent:

- 1.  $T^-$  is a backward Kantorovich operator from C(Y) to USC(X).
- There exists a l.s.c. cost functional c : X × P(Y) → ℝ ∪ {+∞} with σ ↦ c(x, σ) proper and convex for each x ∈ X, and a balayage cone A ∈ LSC(X ⊔ Y) such that

$$U^{-}g(x) := \sup\{\int_{Y} gd\sigma - c(x,\sigma); \sigma \in \mathcal{P}(Y) \text{ and } \delta_{x} \prec_{\mathcal{A}} \sigma\},$$

A gambling house that charges fees: Unlike cost-free gambling houses, a gambler with wealth x, incurs a cost  $c(x, \sigma)$  each time they choose a distribution of gains  $\sigma$ .

## Kantorovich operators are functional capacities

Denote by  $F_b(Y)$  (resp.,  $F^b(X)$ ) the class of functions on Y (resp., X) that are bounded above.

**Theorem:** Let  $T : C(Y) \rightarrow USC(X)$  be a backward Kantorovich operator and let *c* be its cost. Then

1. T can be extended to a map from  $F^b(Y)$  to  $F^b(X)$  via the formula

$$Tg(x) = \sup\{\int_{Y}^{*} gd\nu - c(x,\nu); \nu \in \mathcal{P}(Y)\},\$$

where  $\int_{Y}^{*} g d\nu$  is the outer integral of g with respect to  $\nu$ .

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If T0 is bounded below, then there is a constant k such that T + k is a Choquet functional capacity that maps F<sup>b</sup><sub>+</sub>(Y) to F<sup>b</sup><sub>+</sub>(X).
 If g is a K-analytic function that is bounded on Y, then

$$T^{-}g(x) := \sup\{T^{-}h(x); h \in USC(Y), h \le g\}.$$

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# Kantorovich envelopes

A Choquet functional capacity is a map  $T : F_+(Y) \to F_+(X)$  (The set of all non-negative functions valued in  $\mathbb{R} \cup \{+\infty\}$ ) such that

- 1. T is monotone, i.e.,  $f \leq g \Rightarrow Tf \leq Tg$ .
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**Theorem:** Let  $T : C(Y) \rightarrow USC_b(X)$  be a standard map. Then,

1. (Kantorovich envelope)

$$\underline{T}g(x) := \sup_{\sigma \in \mathcal{P}(Y)} \inf_{h \in C(Y)} \{ \int_{Y} (g-h) \, d\sigma + Th(x) \}$$

is the greatest Kantorovich operator S such that  $S \leq T$  on C(Y).

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2. (Choquet-Kantorovich envelope) If  $T : F_+(Y) \to F_+(X)$  is a functional capacity, then

$$\overline{\mathsf{T}}g(x) := \sup_{\sigma \in \mathcal{P}(Y)} \inf_{O \text{ open}} \{ \int_{Y} (g - \chi_O) \, d\sigma + T \chi_{\bar{O}}(x) \}$$

is the greatest Kantorovich operator S such that  $S(\chi_K) \leq T(\chi_K)$ for every compact K. Let  $T: F_+(Y) \to F_+(X)$  be a functional capacity. Then there exists

- 1. a l.s.c. cost functional  $c : X \times \mathcal{P}(Y) \to \mathbb{R} \cup \{+\infty\}$  with  $\sigma \mapsto c(x, \sigma)$  proper and convex for each  $x \in X$ ,
- 2. a balayage cone  $\mathcal{A} \in LSC(X \sqcup Y)$

such that

$$\underline{\mathrm{T}}g(x) := \sup\{\int_Y gd\sigma - c(x,\sigma); \sigma \in \mathcal{P}(Y) \text{ and } \delta_x \prec_{\mathcal{A}} \sigma\},\$$

and  $\underline{T}$  is the greatest Choquet-Kantorovich functional capacity S such that  $S(\chi_{\kappa}) \leq T(\chi_{\kappa})$  for every compact  $\kappa$ .

#### The secret: Linear Transfers

Let  $\mathcal{T} : \mathcal{P}(X) \times \mathcal{P}(Y) \to \mathbb{R} \cup \{+\infty\}$  be a proper convex and weak<sup>\*</sup> lower semi-continuous on  $\mathcal{M}(X) \times \mathcal{M}(Y)$ . Write  $D(\mathcal{T})$  for its domain.

- For  $\mu \in \mathcal{P}(X)$ , consider the partial maps  $\mathcal{T}_{\mu} : \nu \to \mathcal{T}(\mu, \nu)$  on  $\mathcal{P}(Y)$ ,
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#### The secret: Linear Transfers

Let  $\mathcal{T} : \mathcal{P}(X) \times \mathcal{P}(Y) \to \mathbb{R} \cup \{+\infty\}$  be a proper convex and weak<sup>\*</sup> lower semi-continuous on  $\mathcal{M}(X) \times \mathcal{M}(Y)$ . Write  $D(\mathcal{T})$  for its domain.

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$$\mathcal{T}(\mu,\nu) = \sup\big\{\int_Y g(y)\,d\nu(y) - \int_X T^-g(x)\,d\mu(x);\,g\in C(Y)\big\}.$$

2.  $\mathcal{T}$  is a forward linear transfer, if there exists an operator  $T^+ : C(X) \to LSC(Y)$  such that for each  $\nu \in \mathcal{P}(Y)$ ,

 $\mathcal{T}^*_{\nu}(f) = -\int_Y \mathcal{T}^+(-f)(y) \, d\nu(y) \quad \text{ for any } f \in \mathcal{C}(X).$ 

Hence,

$$\mathcal{T}(\mu,\nu) = \sup \left\{ \int_Y T^+ f(y) \, d\nu(y) - \int_X f(x) \, d\mu(x); \, f \in C(X) \right\}.$$

# Optimal balayage transport

Optimal weak transport (Gozlan et al.) Let  $c : X \times \mathcal{P}(Y) \to \mathbb{R} \cup \{+\infty\}$ be a l.s.c. function such that for each  $x \in X$ , the function  $\sigma \mapsto c(x, \sigma)$  is proper and convex. For  $\mu \in \mathcal{P}(X)$  to  $\nu \in \mathcal{P}(Y)$ , it is

 $\mathcal{T}_{c}(\mu,\nu) := \inf_{\pi} \{ \int_{X} c(x,\pi_{x}) d\mu(x); \pi \in \mathcal{K}(\mu,\nu) \}, \text{where}$ 

$$\mathcal{K}(\mu,\nu) = \left\{ \pi \in \mathcal{P}(X \times Y); \pi_X = \mu, \pi_Y = \nu, \pi(A \times B) = \int_A \pi_x(B) \, d\mu(x) \right\}$$

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Strassen.bis: If  $\mathcal{A}$  is a balayage cone in  $LSC(X \sqcup Y)$  and  $\mu \prec_{\mathcal{A}} \nu$ , then there exists  $\pi \in \mathcal{K}_{\mathcal{A}}(\mu, \nu) = \{\pi \in \mathcal{K}(\mu, \nu); \delta_x \prec_{\mathcal{A}} \pi_x \mid \mu - a.e.\}$ 

Optimal balayage transport

$$\mathcal{B}_{c,\mathcal{A}}(\mu,\nu) = \begin{cases} \inf\{\int_X c(x,\pi_x)d\mu(x) \, ; \, \pi \in \mathcal{K}_{\mathcal{A}}(\mu,\nu)\} & \text{if } \mu \preceq_{\mathcal{A}} \nu, \\ +\infty & \text{otherwise.} \end{cases}$$

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# A seminal duality

**Theorem:** The following are equivalent:

- 1.  $T^-$  is a backward Kantorovich operator from C(Y) to USC(X).
- 2. There is a backward linear transfer  $\mathcal{T} : \mathcal{P}(X) \times \mathcal{P}(Y) \to \mathbb{R} \cup \{+\infty\}$  such that for all  $\mu \in \mathcal{P}(X)$  and  $g \in C(Y)$ ,

$$\mathcal{T}^*_\mu(g) = \int_X T^- g \, d\mu.$$

3. There exists an optimal balayage transport  $\mathcal{T}_{c,\mathcal{A}}$ , which is a backward linear transfer whose Kantorovich operator is given by

$$T^-g(x) := \sup\{\int_Y gd\sigma - c(x,\sigma); \sigma \in \mathcal{P}(Y), \delta_x \prec_{\mathcal{A}} \sigma\}.$$

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**Corollary:** T is positively homogenous iff S is a backward transfer set, i.e,

$$\mathcal{T}(\mu, 
u) = \left\{ egin{array}{cc} 0 & ext{if } (\mu, 
u) \in \mathcal{S} \ +\infty & ext{otherwise}, \end{array} 
ight.$$

is a zero-cost backward linear transfer, in which case

$$\mathcal{S} = \{(\mu, 
u) \in \mathcal{P}(X) imes \mathcal{P}(Y); \ \mu \prec_{\mathcal{A}} 
u\}.$$

#### Proposition (Iterations of Kantorovich operators)

Let  $X_1, ..., X_n$  be *n* compact spaces, and for each i = 1, ..., n,

- $\mathcal{T}_i$  is a backward linear transfer on  $\mathcal{P}(X_{i-1}) \times \mathcal{P}(X_i)$
- $T_i : USC(X_i) \rightarrow USC(X_{i-1})$  is the associated Kantorovich operator. For  $(\mu, \nu) \in \mathcal{P}(X_1) \times \mathcal{P}(X_n)$ , define

 $\mathcal{T}_1 \star \mathcal{T}_2 \dots \star \mathcal{T}_n(\mu, \nu) = \inf \{ \mathcal{T}_1(\mu, \sigma_1) + \mathcal{T}_2(\sigma_1, \sigma_2) \dots + \mathcal{T}_n(\sigma_{n-1}, \nu); \ \sigma_i \in \mathcal{P}(X_i), i = 1, n-1 \}.$ 

Then,  $\mathcal{T} := \mathcal{T}_1 \star \mathcal{T}_2 ... \star \mathcal{T}_n$  is a linear backward transfer with a Kantorovich operator given by

 $T = T_1 \circ T_2 \circ \ldots \circ T_n.$ 

Denote by  $\mathcal{T}_n$  the linear transfer  $\mathcal{T} \star \mathcal{T} \star ... \star \mathcal{T}$  by iterating *n*-times. The corresponding Kantorovich operator is then  $T^n = T \circ T \circ ... \circ T$ .

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• The self-transfer constant of a backward linear transfer  $\mathcal{T}$  and its Kantorovich operator  $\mathcal{T} : C(X) \to USC(X)$  is the -possibly infinite-

$$c(T) := \inf_{\mu \in \mathcal{P}(X)} \sup_{h \in C(X)} \{ \int_X (h - Th) d\mu = \inf_{\mu \in \mathcal{P}(X)} \mathcal{T}(\mu, \mu).$$

• If c(T) is finite, then there exists  $\bar{\mu} \in \mathcal{P}(X)$  such that  $\mathcal{T}(\bar{\mu}, \bar{\mu}) = c(T)$ . Such measures will be called **minimal measures**. • The self-transfer constant of a backward linear transfer  $\mathcal{T}$  and its Kantorovich operator  $\mathcal{T} : C(X) \to USC(X)$  is the -possibly infinite-

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• If c(T) is finite, then there exists  $\bar{\mu} \in \mathcal{P}(X)$  such that  $\mathcal{T}(\bar{\mu}, \bar{\mu}) = c(T)$ . Such measures will be called **minimal measures**.

• A backward subsolution (resp., solution) for T at level  $k \in \mathbb{R}$  is a function  $g \in USC(X)$  so that

1. 
$$Tg + k \leq g$$
 (resp.,  $Tg + k = g$ ) and

2.  $\int_X gd\mu > -\infty$  for some minimal measure  $\mu \in \mathcal{P}(X)$ .

• The self-transfer constant of a backward linear transfer  $\mathcal{T}$  and its Kantorovich operator  $\mathcal{T} : C(X) \to USC(X)$  is the -possibly infinite-

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• The Mañé constant  $c_0$  is the supremum over all  $k \in \mathbb{R}$  such that there exists a subsolution g for T at level k.

#### The self-transfer constant

Let  $\mathcal{T}: \mathcal{P}(X) \times \mathcal{P}(X) \to \mathbb{R} \cup \{+\infty\}$  be a backward linear transfer,  $\mathcal{T}$  its associated backward Kantorovich operator. Then,

1. 
$$c_0 = c(\mathcal{T})$$
  
2.  $c(\mathcal{T}) = \lim_{n \to \infty} \frac{\inf_{(\mu,\nu) \in \mathcal{P}(X) \times \mathcal{P}(X)} \mathcal{T}_n(\mu,\nu)}{n}$ .

3. If  $\overline{\mu}$  is a minimal measure, then for each  $g \in C(X)$ ,

$$\lim_{n\to\infty}\frac{1}{n}\int_X T^n g\,d\bar{\mu}=-c(\mathcal{T}).$$

4. If  $\mathcal{T}$  is bounded above on  $\mathcal{P}(X) \times \mathcal{P}(X)$ , then

$$\frac{\mathcal{T}_n(\mu,\nu)}{n} \to c(\mathcal{T}) \quad \text{uniformly on } \mathcal{P}(X) \times \mathcal{P}(X), \tag{1}$$

and for every  $g \in C(X)$ ,

$$\frac{T^n g(x)}{n} \to -c(T) \quad \text{uniformly on } X. \tag{2}$$

## Weak KAM operators associated to Kantorovich operators

Let  $T : USC(X) \rightarrow USC(X)$  be a backward Kantorovich operator with a finite transfer constant c(T). Say that a Kantorovich operator  $T_{\infty} : USC(X) \rightarrow USC(X)$  is a backward weak KAM operator associated to T if

- 1.  $T_{\infty}$  is idempotent.
- 2.  $TT_{\infty} = T_{\infty}T$ .
- 3.  $T_{\infty}$  maps C(X) to the class of backward weak KAM solutions for T, i.e., for any  $g \in C(X)$ ,

$$TT_{\infty}g+c(T)=T_{\infty}g.$$

The linear transfer associated to  $T_\infty$  is then

$$\mathcal{T}_{\infty}(\mu,\nu) = \sup_{g \in C(X)} \{ \int_X g \, d\nu - \int_X T_{\infty}g \, d\mu \},\,$$

and will be called the Peirls barrier associated to T.

Three cases where we can prove the existence of a weak KAM operator associated to a Kantorovich opeartor:

- 1. When  ${\mathcal T}$  is a weak\*-continuous backward linear transfer.
- 2. When  $c(T) = \inf_{(\mu,\nu)\in\mathcal{P}(X)\times\mathcal{P}(X)} \mathcal{T}(\mu,\nu)$ . For example, when T is 1-positively homogenous.
- 3. When  $\mathcal{T}$  is of bounded oscillation. A linear transfer  $\mathcal{T}$  has bounded oscillation if

$$\limsup_{n\to\infty} \{nc(\mathcal{T}) - \inf_{\mathcal{P}(X)\times\mathcal{P}(X)} \mathcal{T}_n\} < +\infty.$$

For example, when  ${\mathcal T}$  is bounded above.

#### Ergodic properties of continuous linear transfers

Let  $\mathcal{T}$  be a weak\*-continuous backward linear transfer with backward Kantorovich operator  $\mathcal{T}$ . Then, there exists a backward weak KAM operator  $\mathcal{T}_{\infty}^-: \mathcal{C}(X) \to \mathcal{C}(X)$  with a corresponding backward linear transfer  $\mathcal{T}_{\infty}$  so that

1. 
$$(\mathcal{T}_n - nc(\mathcal{T})) \star \mathcal{T}_\infty = \mathcal{T}_\infty = \mathcal{T}_\infty \star (\mathcal{T}_n - nc(\mathcal{T}))$$
 for every  $n \in \mathbb{N}$ .

2. For every  $\mu, \nu \in \mathcal{P}(\mathcal{X})$ , we have

$$\sup\left\{\int_X T_{\infty}^- gd(\nu-\mu); g \in C(X)\right\} \leq \mathcal{T}_{\infty}(\mu,\nu) \leq \liminf_{n \to \infty} (\mathcal{T}_n(\mu,\nu) - nc(\mathcal{T})).$$

3.  $\mathcal{A} := \{ \sigma \in \mathcal{P}(X); \mathcal{T}_{\infty}(\sigma, \sigma) = 0 \}$  contains all minimal measures of  $\mathcal{T}$ .

4. If T is also a forward linear transfer, then there exists conjugate functions  $\psi_0, \psi_1$  for  $T_\infty$  in the sense that

$$\psi_0 = T_\infty^- \psi_1 \quad \psi_1 = T_\infty^+ \psi_0,$$

such that

$$T^-\psi_0+c=\psi_0,\quad T^+\psi_1-c=\psi_1,$$

and

$$\int_X \psi_0 d\mu = \int_X \psi_1 d\mu \,\,$$
 for every  $\mu \in \mathcal{A}.$ 

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## Semi-groups of Kantorovich operators

Let  $\{\mathcal{T}_t\}_{t\geq 0}$  be a family of backward linear transfers on  $\mathcal{P}(X) \times \mathcal{P}(X)$ with associated Kantorovich operators  $\{\mathcal{T}_t\}_{t\geq 0}$ ,

(H0)  $\{\mathcal{T}_t\}_{t\geq 0}$  is a semi-group for inf-convolution:  $\mathcal{T}_{t+s} = \mathcal{T}_t \star \mathcal{T}_s \ (s,t\geq 0)$ 

(H1) For every t > 0, the transfer  $\mathcal{T}_t$  is weak\*-continuous.

(H2) For any  $\epsilon > 0$ ,  $\{\mathcal{T}_t\}_{t \ge \epsilon}$  has common modulus of continuity  $\delta(\epsilon)$ .

**Example:**  $A_t(x, y)$  be a semi-group of equicontinuous cost functions on  $X \times X$ , that is

$$A_{t+s}(x,y) = A_t \star A_s(x,y) := \inf\{A_t(x,z) + A_s(z,y); z \in X\},\$$

and the associated optimal mass transports

$$\mathcal{T}_t(\mu,\nu) = \inf\{\int_{X\times X} A_t(x,y)d\pi(x,y); \pi \in \mathcal{K}(\mu,\nu)\}.$$

 $(\mathcal{T}_t)_t$  is then a semi-group of linear transfers and there is a backward and forward linear transfer  $\mathcal{T}_{\infty}$ , and weak KAM operators  $\mathcal{T}_{\infty}^-$ ,  $\mathcal{T}_{\infty}^+$  such that:

#### Weak KAM theories

If 
$$c := c((\mathcal{T}_t)_t) := \lim_{t \to \infty} \frac{\inf_{\mu,\nu \in \mathcal{P}(X)} \mathcal{T}_t(\mu,\mu)}{t}$$
. Then  
1.  $c = \min\{\int_{X \times X} A_1(x, y) d\pi; \pi \in \mathcal{P}(X \times X), \pi_1 = \pi_2\}$   
2.  $A_{\infty}(x, y) := \liminf_{t \to \infty} (A_t(x, y) - ct)$  is continuous on  $X \times X$ , and  
•  $\mathcal{T}_{\infty}(\mu, \nu) = \mathcal{T}_{A_{\infty}}(\mu, \nu) := \inf\{\int_{X \times X} A_{\infty}(x, y) d\pi(x, y); \pi \in \mathcal{K}(\mu, \nu)\},$   
•  $\mathcal{T}_{\infty}^- f(x) = \sup\{f(y) - A_{\infty}(x, y); y \in X\}, \ \mathcal{T}_{\infty}^+ f(y) = \inf\{f(x) + A_{\infty}(x, y); x \in X\}.$ 

3. The minimizing measures in (1) are all supported on the set

$$D := \{(x, y) \in X \times X ; A_1(x, y) + A_{\infty}(y, x) = c\}.$$

4. There exists conjugate functions  $u^-, u^+$  for  $\mathcal{T}_\infty$  in the sense that

$$\begin{split} u^{-}(x) &= \sup\{u^{+}(y) - A_{\infty}(x, y); \ y \in X\}, \quad u^{+}(y) = \inf\{u^{-}(x) + A_{\infty}(x, y); \ x \in X\}, \\ T_{t}^{-}u^{-} + ct &= u^{-}, \quad T_{t}^{+}u^{+} - ct = u^{+}) \ \text{for all } t \geq 0. \\ u^{-}(x) &= u^{+}(x) \ \text{whenever } A_{\infty}(x, x) = 0. \end{split}$$

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## Weak KAM solutions in Lagrangian dynamics

Let L be a time-independent Tonelli Lagrangian on a compact Riemanian manifold M, and consider  $T_t$  to be the cost minimizing transport

$$\mathcal{T}_t(\mu,\nu) = \inf\{\int_{M \times M} A_t(x,y) d\pi(x,y); \pi \in \mathcal{K}(\mu,\nu)\}, \text{where}$$

$$A_t(x,y) := \inf\{\int_0^t L(\gamma(s),\dot{\gamma}(s))ds \, ; \, \gamma \in C^1([0,t];M); \gamma(0) = x, \gamma(t) = y\}.$$

The backward (forward) Lax-Oleinik semi-group is defined for t > 0, via

$$S_t^-u(x) = \sup\{u(\gamma(t)) - \int_0^t L(\gamma(s), \dot{\gamma}(s)) ds; \gamma \in C^1([0, t]; M), \gamma(0) = x\},$$

$$S_t^+ u(x) := \inf \{ u(\gamma(0)) + \int_0^t L(\gamma(s), \dot{\gamma}(s)) ds \, ; \, \gamma \in C^1([0, t]; M), \gamma(t) = x \}.$$

A function  $u \in C(M)$  is said to be a *backward (resp., forward) weak* KAM solution if  $S_t^-u + ct = u$  (resp.,  $S_t^+u - ct = u$ ) for all  $t \ge 0$ . **Theorem:** There exists a unique constant  $c \in \mathbb{R}$  such that:

- 1. (Fathi) There exists weak KAM solutions, i.e.,  $u_-: M \to \mathbb{R}$  (resp.  $u_+$ ) such that  $S_t^- u_- + ct = u_-$  (resp.  $S_t^+ u_- ct = u_-$ ) for  $t \ge 0$ .
- (Bernard-Buffoni) Let A<sub>∞</sub>(x, y) := lim inf<sub>t→∞</sub> A<sub>t</sub>(x, y) tc denotes the Peierls barrier function. Then,

$$\inf\{\int_{M\times M}A_{\infty}(x,y)d\pi(x,y); \pi\in\mathcal{K}(\mu,\nu)\}=\sup_{u_+,u_-}\{\int_{M}u_+d\nu-\int_{M}u_-d\mu\},$$

and

$$u_+ = u_- \text{ on } \mathcal{A} := \{x \in M ; A_\infty(x, x) = 0\}.$$

 (Bernard-Buffoni) c = min<sub>π</sub> ∫<sub>M×M</sub> A<sub>1</sub>(x, y)dπ(x, y), over all π ∈ P(M × M) with equal first and second marginals. The minimizing measures are all supported on

$$\mathcal{D}:=\{(x,y)\in M\times M\,;\,A_1(x,y)+A_\infty(y,x)=c\}.$$

- 4. (Mather)  $c = \inf_m \int_{TM} L(x, v) dm(x, v)$  over all measures  $m \in \mathcal{P}(TM)$  which are invariant under the Euler-Lagrange flow.
- 5. (Fathi) A continuous function  $u: M \to \mathbb{R}$  is a viscosity solution of

$$H(x,\nabla u(x))=c$$

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if and only if it is Lipschitz and u is a backward weak KAM solution.

## Stochastic (2d order Fathi-Mather theory

(Gomez, Mikami)  $M = \mathbb{T}^d := \mathbb{R}^d / \mathbb{Z}^d$  being the *d*-dimensional flat torus,  $(\Omega, \mathcal{F}, \mathcal{P})$  a complete probability space with normal filtration  $\{\mathcal{F}_t\}_{t\geq 0}$ , Let  $\mathcal{A}_{[0,t]}$  be the set of continuous semi-martingales  $X : \Omega \times [0, t] \to M$  such that for some drift  $\beta_X : [0, t] \times C([0, t]) \to \mathbb{R}^d$ ,

$$dX_t = \beta(t, X)dt + dW_t$$

where  $W_t$  is an *M*-valued Brownian motion.

$$\mathcal{T}_t(\mu,\nu) := \inf \left\{ \mathbb{E} \int_0^t L(X(s),\beta_X(s,X)) \varsigma; X(0) \sim \mu, X(t) \sim \nu, X \in \mathcal{A}_{[0,t]} \right\},\$$

is then a backward linear transfer, with Kantorovich operator

$$T_t f(x) := \sup_{X \in \mathcal{A}_{[0,t]}} \left\{ \mathbb{E}\left[ f(X(t)) - \int_0^t L(X(s), \beta_X(s, X)) \$ | X(0) = x \right] \right\}.$$

$$\mathcal{N}_0 := \left\{ m \in \mathcal{P}(TM); \int_{TM} \left[ \frac{1}{2} \Delta(x) \phi + v \cdot \nabla \phi(x) \right] dm(x, v) = 0 \text{ for all } \phi \in C^2(M) \right\}.$$

(Euler-Lagrange Flow invariant measures on phase space)

- 1.  $c := \inf \{ \mathcal{T}_1(\mu, \mu) ; \mu \in \mathcal{P}(M) \} = \inf \{ \int_{\mathcal{T}M} L(x, v) m(x, v); m \in \mathcal{N}_0 \}.$ Infimum is attained by a measure  $\overline{m}$ , a stochastic Mather measure. Its projection  $\mu_{\overline{m}}$  on  $\mathcal{P}(M)$  is a minimiser for  $\mathcal{T}_1$ .
- 2. There exists backward weak KAM solutions  $T_t u + ct = u$  for  $t \ge 0$ ,  $u \in C(M)$ ,
- 3. The backward weak KAM solutions are exactly the viscosity solutions of the stationary Hamilton-Jacobi-Bellman equation  $\frac{1}{2}\Delta u + H(x, D_x u) = c$ .

# Symbolic dynamics (Garibaldi-Lopez)

Let M is an  $r \times r$  transition matrix, whose  $\{0,1\}$  entries specify allowable transitions.

$$\Sigma = \{x \in \{1, ..., r\}^{\mathbb{N}} ; M(x_i, x_{i+1}) = 1, \forall i \geq 0\}$$

the set of admissible words, and its dual

$$\Sigma^* = \{y \in \{1, ..., r\}^{\mathbb{N}} ; M(y_{i+1}, y_i) = 1, \forall i \ge 0\}$$

and consider

$$\hat{\Sigma} = \{(y, x) \in \Sigma^* \times \Sigma; M(y_0, x_0) = 1\}.$$

Assume  $\Sigma_x^* := \{y \in \Sigma^* ; (y, x) \in \hat{\Sigma}\} \neq \emptyset, \forall x \in \Sigma.$ Consider the time-evolution map  $\sigma : \Sigma \to \Sigma$  and  $\tau : \hat{\Sigma} \to \Sigma$  defined as

$$\sigma(x_0, x_1, ...) = (x_1, x_2, ...)$$
 and  $\tau(y, x) = (y_0, x_0, x_1, ...).$ 

The set of holonomic probability measures is

$$\mathcal{M}_0(\hat{\Sigma}) := \left\{ \mu \in \mathcal{P}(\hat{\Sigma}) \ ; \ \int_{\hat{\Sigma}} f(\tau_y(x)) - f(x) \ d\mu(y, x) = 0 \right\}.$$

**Theorem:** Given  $A \in C(\hat{\Sigma})$ , define  $\beta_A := \max_{\hat{\mu} \in \mathcal{M}_0(\hat{\Sigma})} \int_{\hat{\Sigma}} A(y, x) d\hat{\mu}(y, x)$ . Then

$$\beta_A = \inf_{f \in C(\Sigma)} \max_{(y,x) \in \hat{\Sigma}} \{A(y,x) + f(x) - f(\tau_y(x))\}.$$

There exists  $u \in USC(\Sigma)$  such that

$$\inf_{y\in\Sigma_x^*} \{u(\tau_y(x)) - A(y,x) + \beta(A)\} = u(x) \quad \forall x \in \Sigma.$$

Thank you

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