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Optimal mass transportation and Mather theory

Patrick Bernard and Boris Buffoni

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Abstract: We study the Monge transportation problem when the cost is the action associated to a Lagrangian function on a compact manifold. We show that the transportation can be interpolated by a Lipschitz lamination. We describe several direct variational problems the minimizers of which are these Lipschitz laminations. We prove the existence of an optimal transport map when the transported measure is absolutely continuous. We explain the relations with Mather's minimal measures.

Résumé: On étudie le problème de transport de Monge lorsque le cout est l'action associée à un Lagrangien sur une variété compacte. On montre que le transport peut être interpolé par une lamination lipschitzienne. On décrit plusieurs problèmes variationnels directs dont ces laminations sont les minimiseurs. On montre l'existence d'une application de transport optimale lorsque la mesure transportée est absolument continue. On explique les relations avec les mesures minimisantes de Mather.

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Boris Buffoni School of Mathematics École Polytechnique Fédérale-Lausanne SB/IACS/ANA Station 8 1015 Lausanne Switzerland boris.buffoni@epfl.ch Several observations have recently renewed the interest for the classical topic of optimal mass transportation, whose primary origin is attributed to Monge a few years before French revolution. The framework is as follows. A space M is given, which in the present paper will be a compact manifold, as well as a continuous cost function $c(x,y): M \times M \longrightarrow \mathbb{R}$. Given two probability measures μ_0 and μ_1 on M, the mappings $\Psi: M \longrightarrow M$ which transport μ_0 into μ_1 and minimize the total cost $\int_M c(x, \Psi(x)) d\mu_0$ are studied. It turns out, and it was the core of the investigations of Monge, that these mappings have very remarkable geometric properties, at least at a formal level.

Only much more recently was the question of the existence of optimal objects rigorously solved by Kantorovich in a famous paper of 1942. Here we speak of optimal objects, and not of optimal mappings, because the question of existence of an optimal mapping is ill-posed, so that the notion of optimal objects has to be relaxed, in a way that nowadays seems very natural, and that was discovered by Kantorovich.

Our purpose here is to continue the work initiated by Monge, recently awakened by Brenier and enriched by other authors, on the study of geometric properties of optimal objects. The costs functions we consider are natural generalizations of the cost $c(x,y) = d(x,y)^2$ considered by Brenier and many other authors. The book [39] gives some ideas of the applications expected from this kind of questions. More precisely, we consider a Lagrangian function L(x,v,t): $TM \times \mathbb{R} \longrightarrow \mathbb{R}$ which is convex in v and satisfies standard hypotheses recalled later, and define our cost by

$$c(x,y) = \min_{\gamma} \int_{0}^{1} L(\gamma(t), \dot{\gamma}(t), t) dt$$

where the minimum is taken on the set of curves $\gamma:[0,1] \longrightarrow M$ satisfying $\gamma(0)=x$ and $\gamma(1)=y$. Note that this class of costs does not contain the very natural cost c(x,y)=d(x,y). Such costs are studied in a second paper [9].

Our main result is that the optimal transports can be interpolated by measured Lipschitz laminations, or geometric currents in the sense of Ruelle and Sullivan. Interpolations of transport have already been considered by Benamou, Brenier and McCann for less general cost functions, and with different purposes. Our methods are inspired by the theory of Mather, Mañé and Fathi on Lagrangian dynamics, and we will detail rigorously the relations between these theories. Roughly, they are exactly similar except that mass transportation is a Dirichlet boundary value problem, while Mather theory is a periodic boundary value Problem. We will also prove, extending works of Brenier, Gangbo, McCann, Carlier, and other authors, that the optimal transportation can be performed by a Borel map with the additional assumption that the transported measure is absolutely continuous.

Various connections between Mather-Fathi theory, optimal mass transportation and Hamilton-Jacobi equations have recently been discussed, mainly at a formal level, in the literature, see for example [39], or [19], where they are all presented as infinite dimensional linear programming problems. This have motivated a lot of activity around the interface between Aubry-Mather theory and optimal transportation, some of which overlap partly the present work. For example, at the moment of submitting the paper, we have been informed of the existence of the recent preprints of De Pascale, Gelli and Granieri, [15], and of Granieri, [26]. We had also been aware of a manuscript by Wolansky [40] for a few weeks, which, independently, and by somewhat different methods, obtains results similar to ours. Note however that Lipschitz regularity, which we consider as one of our most important results, was not obtained in this preliminary version of [40]. It is worth also mentioning the papers [36] of Pratelli and [31] of Loeper.

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1 Introduction

We present the context and the main results of the paper.

1.1 Lagrangian, Hamiltonian and cost

In all the present paper, the space M will be a compact and connected Riemannian manifold without boundary. Some standing notations are gathered in the appendix. Let us fix a positive real number T, and a Lagrangian function

$$L \in C^2(TM \times [0, T], \mathbb{R}).$$

A curve $\gamma \in C^2([0,T],M)$ is called an extremal if it is a critical point of the action

$$\int_0^T L(\gamma(t), \dot{\gamma}(t), t) dt$$

with fixed endpoints. It is called a minimizing extremal if it is minimizing the action. We assume:

convexity For each $(x,t) \in M \times [0,T]$, the function $v \mapsto L(x,v,t)$ is convex with positive definite Hessian at each point.

superlinearity For each $(x,t) \in M \times [0,T]$, we have $L(x,v,t)/\|v\| \longrightarrow \infty$ as $\|v\| \longrightarrow \infty$. Arguing as in [20] (Lemma 3.2.2), this implies that for all $\alpha > 0$ there exists C > 0 such that $L(x,v,t) \geqslant \alpha \|v\| - C$ for all $(x,v,t) \in TM \times [0,T]$.

completeness For each $(x, v, t) \in TM \times [0, T]$, there exists one and only one extremal $\gamma \in C^2([0, T], M)$ such that $(\gamma(t), \dot{\gamma}(t)) = (x, v)$.

We associate to the Lagrangian L a Hamiltonian function $H \in C^2(T^*M \times [0,T],\mathbb{R})$ given by

$$H(x, p, t) = \max_{v} p(v) - L(x, v, t).$$

We endow the cotangent bundle T^*M with its canonical symplectic structure, and associate to the Hamiltonian H the time-dependent vectorfield Y on T^*M , which is given by

$$Y = (\partial_n H, -\partial_x H)$$

in any canonical local trivialisation of T^*M . The hypotheses on L can be expressed in terms of the function H:

convexity For each $(x,t) \in M \times [0,T]$, the function $p \longmapsto H(x,p,t)$ is convex with positive definite Hessian at each point.

superlinearity For each $(x,t) \in M \times [0,T]$, we have $H(x,p,t)/\|p\| \longrightarrow \infty$ as $\|p\| \longrightarrow \infty$.

completeness Each solution of the equation $(\dot{x}(t), \dot{p}(t)) = Y(x(t), p(t), t)$ can be extended to the interval [0, T]. We can then define, for all $s, t \in [0, T]$, the flow φ_s^t of Y from times s to time t.

In addition, the mapping $\partial_v L: TM \times [0,T] \longrightarrow T^*M \times [0,T]$ is a C^1 diffeomorphism, whose inverse is the mapping $\partial_p H$. These diffeomorphisms conjugate Y with a time-dependent vectorfield E on TM. We denote the flow of E by $\psi_s^t: TM \longrightarrow TM$ $(s,t \in [0,T])$, which is such that $\psi_s^s = Id$ and $\partial_t \psi_s^t = E_t \circ \psi_s^t$, where as usual E_t denotes the vectorfield E(.,t) on TM. The diffeomorphisms $\partial_v L$ and $\partial_p H$ conjugate the flows ψ_s^t and φ_s^t . Moreover the extremals are the projection of the integral curves of E and

$$\left(\pi \circ \psi_s^t, \, \partial_t(\pi \circ \psi_s^t)\right) = \psi_s^t,\tag{1}$$

where $\pi:TM\to M$ is the canonical projection. In (1), $\partial_t(\pi\circ\psi_s^t)$ is seen as a vector in the tangent space of M at $\pi\circ\psi_s^t$. If $\partial_t(\pi\circ\psi_s^t)$ is seen as a point in TM, (1) becomes simply $\partial_t(\pi\circ\psi_s^t)=\psi_s^t$.

For each $0 \le s < t \le T$, we define the cost function

$$c_s^t(x,y) = \min_{\gamma} \int_s^t L(\gamma(\sigma), \dot{\gamma}(\sigma), \sigma) d\sigma$$

where the minimum is taken on the set of curves $\gamma \in C^2([s,t],M)$ satisfying $\gamma(s) = x$ and $\gamma(t) = y$. That this minimum exists is a standard result under our hypotheses, see [33] or [20].

Proposition 1. Let us fix a subinterval $[s,t] \subset [0,T]$. The set $\mathcal{E} \subset C^2([s,t],M)$ of minimizing extremals is compact for the C^2 topology.

Let us mention that, for each $(x_0, s) \in M \times [0, T]$, the function $(x, t) \longmapsto c_s^t(x_0, x)$ is a viscosity solution of the Hamilton-Jacobi equation

$$\partial_t u + H(x, \partial_x u, t) = 0$$

on $M \times]s, T[$. This remark may help the reader in understanding the key role which will be played by this equation in the sequel.

1.2 Monge-Kantorovich theory

We recall the basics of Monge-Kantorovich duality. The proofs are available in many texts on the subjects, for example [1, 37, 39]. We assume that M is a compact manifold and that c(x, y) is a continuous cost function on $M \times M$, which will later be one of the costs c_s^t defined above. Given two Borel probability measures μ_0 and μ_1 on M, a transport plan between μ_0 and μ_1 is a measure on $M \times M$ which satisfies

$$\pi_{0t}(\eta) = \mu_0 \text{ and } \pi_{1t}(\eta) = \mu_1,$$

where $\pi_0: M \times M \longrightarrow M$ is the projection on the first factor, and π_1 is the projection on the second factor. We denote by $\mathcal{K}(\mu_0, \mu_1)$, after Kantorovich, the set of transport plans. Kantorovich proved the existence of a minimum in the expression

$$C(\mu_0, \mu_1) = \min_{\eta \in \mathcal{K}(\mu_0, \mu_1)} \int_{M \times M} c d\eta$$

for each pair (μ_0, μ_1) of probability measures on M. Here we will denote by

$$C_s^t(\mu_0, \mu_1) := \min_{\eta \in \mathcal{K}(\mu_0, \mu_1)} \int_{M \times M} c_s^t(x, y) d\eta(x, y)$$
 (2)

the optimal value associated to our family of costs c_s^t The plans which realize this minimum are called optimal transfer plans. A pair (ϕ_0, ϕ_1) of continuous functions is called an admissible Kantorovich pair if is satisfies the relations

$$\phi_1(x) = \min_{y \in M} \phi_0(y) + c(y, x)$$
 and $\phi_0(x) = \max_{y \in M} \phi_1(y) - c(x, y)$

for all point $x \in M$. Note that the admissible pairs are composed of Lipschitz functions if the cost c is Lipschitz, which is the case of the costs c_s^t when s < t. Another discovery of Kantorovich is that

$$C(\mu_0, \mu_1) = \max_{\phi_0, \phi_1} \left(\int_M \phi_1 d\mu_1 - \int_M \phi_0 d\mu_0 \right)$$
 (3)

where the maximum is taken on the set of admissible Kantorovich pairs (ϕ_0, ϕ_1) . This maximization problem is called the dual Kantorovich problem, the admissible pairs which reach this maximum are called optimal Kantorovich pairs. The direct problem (2) and dual problem (3) are related as follows.

Proposition 2. If η is an optimal transfer plan, and if (ϕ_0, ϕ_1) is a Kantorovich optimal pair, then the support of η is contained in the set

$$\{(x,y)\in M^2 \text{ such that } \phi_1(y)-\phi_0(x)=c(x,y)\}\subset M\times M.$$

Let us remark that the knowledge of the set of Kantorovich admissible pairs is equivalent to the knowledge of the cost function c.

Lemma 3. We have

$$c(x,y) = \max_{(\phi_0,\phi_1)} \phi_1(y) - \phi_0(x)$$

where the maximum is taken on the set of Kantorovich admissible pairs.

PROOF. This maximum is clearly less that c(x,y). For the other inequality, let us fix points x_0 and y_0 in M, and consider the functions $\phi_1(y) = c(x_0, y)$ and $\phi_0(x) = \max_{y \in M} \phi_1(y) - c(x, y)$. We have $\phi_1(y_0) - \phi_0(x_0) = c(x_0, y_0) - 0 = c(x_0, y_0)$. So it is enough to prove that the pair (ϕ_0, ϕ_1) is an admissible Kantorovich pair, and more precisely that $\phi_1(y) = \min_{x \in M} \phi_0(x) + c(x, y)$. We have

$$\phi_0(x) + c(x,y) \geqslant c(x_0,y) - c(x,y) + c(x,y) \geqslant c(x_0,y) = \phi_1(y)$$

which gives the inequality $\phi_1(y) \leq \min_{x \in M} \phi_0(x) + c(x,y)$. On the other hand, we have

$$\min_{x \in M} \phi_0(x) + c(x, y) \leqslant \phi_0(x_0) + c(x_0, y) = c(x_0, y) = \phi_1(y).$$

1.3 Interpolations

In this section, the Lagrangian L and time T > 0 are fixed. It is not hard to see that, if μ_1, μ_2 and μ_3 are three probability measures on M, and if $t_1 \le t_2 \le t_3 \in [0, T]$ are three times, then we have the inequality

$$C_{t_1}^{t_3}(\mu_1, \mu_3) \leqslant C_{t_1}^{t_2}(\mu_1, \mu_2) + C_{t_2}^{t_3}(\mu_2, \mu_3).$$

The family $\mu_t, t \in [0, T]$ of probability measures on M is called an interpolation between μ_0 and μ_T if it satisfies the equality

$$C_{t_1}^{t_3}(\mu_{t_1}, \mu_{t_3}) = C_{t_1}^{t_2}(\mu_{t_1}, \mu_{t_2}) + C_{t_2}^{t_3}(\mu_{t_2}, \mu_{t_3})$$

for all $0 \le t_1 \le t_2 \le t_3 \le T$. Our main result is the following:

Theorem A. For each pair μ_0, μ_T of probability measures, there exist interpolations between μ_0 and μ_T . Moreover, each interpolation $(\mu_t), t \in [0, T]$ is given by a Lipschitz measured lamination in the following sense:

Eulerian description: There exists a bounded locally Lipschitz vectorfield $X(x,t): M \times]0, T[\longrightarrow TM \text{ such that, if } \Psi^t_s, (s,t) \in]0, T[^2 \text{ is the flow of } X \text{ from time } s \text{ to time } t, \text{ then } (\Psi^t_s)_{\sharp} \mu_s = \mu_t \text{ for each } (s,t) \in]0, T[^2.$

Lagrangian description : There exists a family $\mathcal{F} \subset C^2([0,T],M)$ of minimizing extremals γ of L, which is such that the relation $\dot{\gamma}(t) = X(\gamma(t),t)$ holds for each $t \in]0,T[$ and for each $\gamma \in \mathcal{F}$. The set

$$\tilde{\mathcal{T}} = \{(\gamma(t), \dot{\gamma}(t), t), t \in]0, T[, \gamma \in \mathcal{F}\} \subset TM \times]0, T[$$

is invariant under the Euler-Lagrange flow ψ . The measure μ_t is supported on $\mathcal{T}_t = \{\gamma(t), \gamma \in \mathcal{F}\}$. In addition, there exists a continuous family $m_t, t \in [0,T]$ of probability measures on TM such that m_t is concentrated on $\tilde{\mathcal{T}}_t = \{(\gamma(t), \dot{\gamma}(t)), \gamma \in \mathcal{F}\}$ for each $t \in [0,T]$, such that $\pi_{\sharp} m_t = \mu_t$ for each $t \in [0,T]$, and such that

$$m_t = (\psi_s^t)_{\sharp} m_s$$

for all $(s,t) \in [0,T]^2$.

Hamilton-Jacobi equation : There exists a Lipschitz and C^1 function $v(x,t): M \times]0, T[\longrightarrow \mathbb{R}$ which satisfies the inequation

$$\partial_t v + H(x, \partial_x v, t) \leq 0$$

with equality if and only if $(x,t) \in \mathcal{T} = \{(\gamma(t),t), \gamma \in \mathcal{F}, t \in]0,T[\}$, and such that $X(x,t) = \partial_{\nu}H(x,\partial_{x}v(x,t),t)$ for each $(x,t) \in \mathcal{T}$.

Uniqueness: There may exist several different interpolations. However, it is possible to choose the vectorfield X, the family \mathcal{F} and the sub-solution v in such a way that the statements above hold for all interpolations μ_t with these fixed X, \mathcal{F} and v. For each $s < t \in]0,T[$, the measure $(Id \times \Psi^t_s)_{\sharp}\mu_s$ is the only optimal transport plan in $\mathcal{K}(\mu_s,\mu_t)$ for the cost c_s^t . This implies that

$$\int_{M} c_s^t(x, \Psi_s^t(x)) d\mu_s(x) = C_s^t(\mu_s, \mu_t).$$

Let us comment a bit the preceding statement. The set $\tilde{\mathcal{T}} \subset TM \times]0,T[$ is the image by the Lipschitz map $(x,t) \to (X(x,t),t)$ of the set $\mathcal{T} \subset TM \times]0,T[$. We shall not take $X(x,t) = \partial_p H(x,\partial_x v(x,t),t)$ outside of \mathcal{T} because we do not prove that this vectorfield is Lipschitz outside of \mathcal{T} . The data of the vectorfield X outside of \mathcal{T} is immaterial: any Lipschitz extension of $X_{|\mathcal{T}}$ will fit. Note also that the relation

$$\Psi_s^t = \pi \circ \psi_s^t \circ X_s \tag{4}$$

holds on \mathcal{T}_s , where $X_s(.) = X(.,s)$.

The vectorfield X in the statement depends on the transported measures μ_0 and μ_T . The Lipschitz constant of X, however, can be fixed independently of these measures, as we now state (see Proposition 13, Proposition 19, Theorem 3 and (11)):

Addendum There exists a decreasing function $K(\epsilon):]0, T/2[\longrightarrow]0, \infty[$, which depends only on the time T and on the Lagrangian L, and such that, for each pair μ_0, μ_T of probability measures, one can choose the vectorfield X in Theorem A in such a way that X is $K(\epsilon)$ -Lipschitz on $M \times [\epsilon, T - \epsilon]$ for each $\epsilon \in]0, T/2[$.

Proving Theorem A is the main goal of the present paper. We will present in section 2 some direct variational problems which are well-posed and of which the transport interpolations are in some sense the solutions. We believe that these variational problems are interesting in themselves. In order to describe the solutions of the variational problem, we will rely on a dual approach based on the Hamilton-Jacobi equation, inspired from Fathi's approach to Mather theory, as detailed in section 3. The solutions of the problems of section 2, as well as the transport interpolations, are then described in section 4, which ends the proof of Theorem A.

1.4 Case of an absolutely continuous measure μ_0

Additional conclusions concerning optimal transport can usually be obtained when the initial measure μ_0 is absolutely continuous. For example a standard question is whether the optimal transport can be realized by an optimal mapping.

A transport map is a Borel map $\Psi: M \longrightarrow M$ which satisfies $\Psi_{\sharp}\mu_0 = \mu_1$. To any transport map Ψ is naturally associated the transport plan $(Id \times \Psi)_{\sharp}\mu_0$, called the induced transport plan. An optimal map is a transport map $\Psi: M \longrightarrow M$ such that

$$\int_{M} c_{T}(x, \Psi(x)) d\mu_{0} \leqslant \int_{M} c_{T}(x, F(x)) d\mu_{0}$$

for any transport map F. It turns out that, under the assumption that μ_0 has no atoms, a transport map is optimal if and only if the induced transport plan is an optimal transport plan, see [1], Theorem 2.1. In other words, we have

$$\inf_{\Psi} \int_{M} c(x, \Psi(x)) d\mu_{0}(x) = C(\mu_{0}, \mu_{1}),$$

where the infimum is taken on the set of transport maps from μ_0 to μ_1 . This is a general result which holds for any continuous cost c. It is a standard question, which turns out to be very hard for certain cost functions, whether the infimum above is reached, or in other words whether there exists an optimal transport plan which is induced from a transport map. Part of the result below is that this holds true in the case of the cost c_0^T . The method we use to prove this is an elaboration on ideas due to Brenier, see [12] and developed for instance in [24], (see also [23]) and [16], which is certainly the closest to our needs.

Theorem B. Assume that μ_0 is absolutely continuous with respect to the Lebesgue class on M. Then for each final measure μ_T , there exists one and only one interpolation $\mu_t, t \in [0,T]$, and each interpolating measure $\mu_t, t < T$ is absolutely continuous. In addition, there exists a family $\Psi_0^t: M \to M, t \in]0,T]$, of Borel maps such that $(Id \times \Psi_0^t)_{\sharp}\mu_0$ is the only optimal transfer plan in $\mathcal{K}(\mu_0,\mu_t)$ for the cost function c_0^t . Consequently, we have

$$\int_{M} c_0^t(x, \Psi_0^t(x)) d\mu_0(x) = C_0^t(\mu_0, \mu_t), \quad 0 < t \leqslant T.$$

If μ_T , instead of μ_0 , is assumed absolutely continuous, then there exists one and only one interpolation, and each interpolating measure μ_t , $t \in]0,T]$ is absolutely continuous.

This theorem will be proved and commented in section 5.

1.5 Mather theory

Let us now assume that the Lagrangian function is defined for all times, $L \in C^2(TM \times \mathbb{R}, \mathbb{R})$ and, in addition to the standing hypotheses, satisfies the periodicity condition

$$L(x, v, t + 1) = L(x, v, t)$$

for all $(x, v, t) \in TM \times \mathbb{R}$. A Mather measure, see [33], is a compactly supported probability measure m_0 on TM which is invariant in the sense that $(\psi_0^1)_{\sharp}m_0 = m_0$ and is minimizing the action

$$A_0^1(m_0) = \int_{TM \times [0,1]} L(\psi_0^t(x,v), t) dm_0 dt.$$

The major discovery of [33] is that Mather measures are supported on the graph of a Lipschitz vectorfield. Let us call α the action of Mather measures –this number is the value at zero of the α function defined by Mather in [33]. Let us now explain how this theory of Mather is related to, and can be recovered from, the content of our paper.

Theorem C. We have

$$\alpha = \min_{\mu} C_0^1(\mu, \mu),$$

where the minimum is taken on the set of probability measures on M. The mapping $m_0 \mapsto \pi_{\sharp} m_0$ is a bijection between the set of Mather measures m_0 and the set of probability measures μ on M satisfying $C_0^1(\mu,\mu) = \alpha$. There exists a Lipschitz vectorfield X_0 on M such that all the Mather measures are supported on the graph of X_0 .

This theorem will be proved in section 6, where the bijection between Mather measures and measures minimizing $C_0^1(\mu,\mu)$ will be precised.

2 Direct variational problems

We state two different variational problems whose solutions are the interpolated transports. We believe that these problems are interesting on their own. They will also be used to prove Theorem A.

2.1 Measures

This formulation parallels Mather's theory. It can also be related to the generalized curves of L. C. Young. Let μ_0 and μ_T be two probability Borel measures on M. Let $m_0 \in \mathcal{B}_1(TM)$ be a Borel probability measure on the tangent bundle TM. We say that m_0 is an initial transport measure if the measure η on $M \times M$ given by

$$\eta = (\pi \times (\pi \circ \psi_0^T))_{\sharp} m_0$$

is a transport plan, where $\pi: TM \longrightarrow M$ is the canonical projection. We call $\mathcal{I}(\mu_0, \mu_T)$ the set of initial transport measures. To an initial transport measure m_0 , we associate the continuous family of measures

$$m_t = (\psi_0^t)_{\sharp} m_0, t \in [0, T]$$

on TM, and the measure m on $TM \times [0,T]$ given by

$$m = m_t \otimes dt = ((\psi_0^t)_{\sharp} m_0) \otimes dt.$$

Note that the linear mapping $m_0 \longmapsto m = ((\psi_0^t)_{\sharp} m_0) \otimes dt$ is continuous from $\mathcal{B}(TM)$ to $\mathcal{B}(TM \times [0,T])$ endowed with the weak topology, see appendix.

Lemma 4. The measure m satisfies the relation

$$\int_{TM\times[0,T]} \partial_t f(x,t) + \partial_x f(x,t) \cdot v \, dm(x,v,t) = \int_M f_T d\mu_T - \int_M f_0 d\mu_0 \tag{5}$$

for each function $f \in C^1(M \times [0,T], \mathbb{R})$, where f_t denotes the function $x \longmapsto f(x,t)$.

PROOF. Setting $\widetilde{f}(x,v,t) = f(x,t)$, $g_1(x,v,t) = \partial_t f(x,t) = \partial_t \widetilde{f}(x,v,t)$ and $g_2(x,v,t) = \partial_x f(x,t) \cdot v$, we have

$$\int_{TM\times[0,T]} \partial_t f(x,t) + \partial_x f(x,t) \cdot v \, dm(x,v,t) = \int_0^T \int_{TM} (g_1 + g_2) \circ \psi_0^t \, dm_0 dt.$$

Noticing that, in view of equation (1), we have

$$\partial_t(\widetilde{f} \circ \psi_0^t) = g_1 \circ \psi_0^t + g_2 \circ \psi_0^t$$

we obtain that

$$\int_{TM\times[0,T]} \partial_t f(x,t) + \partial_x f(x,t) \cdot v \, dm(x,v,t) = \int_{TM} (\widetilde{f} \circ \psi_0^T - \widetilde{f}) dm_0 = \int_M f_T d\mu_T - \int_M f_0 d\mu_0$$

as desired.

Definition 5. A finite Borel measure on $TM \times [0,T]$ which satisfies (5) is called a transport measure. We denote by $\mathcal{M}(\mu_0, \mu_T)$ the set of transport measures. A transport measure which is induced from an initial measure m_0 is called an invariant transport measure. The action of the transport measure m is defined by

$$A(m) = \int_{TM \times [0,T]} L(x,v,t) dm \in \mathbb{R} \cup \{\infty\}$$

The action $A(m_0)$ of an initial transport measure is defined as the action of the associated transport measure m. We will also denote this action by $A_0^T(m_0)$ when we want to insist on the time interval. We have

$$A_0^T(m_0) = \int_{TM \times [0,T]} L(\psi_0^t(x,v), t) dm_0 dt.$$

Notice that initial tranport measures exist:

Proposition 6. The mapping $(\pi \times (\pi \circ \psi_0^T))_{\sharp} : \mathcal{I}(\mu_0, \mu_T) \longrightarrow \mathcal{K}(\mu_0, \mu_T)$ is surjective. In addition, for each transport plan η , there exists a compactly supported initial transport measure m_0 such that $(\pi \times (\pi \circ \psi_0^T))_{\sharp} m_0 = \eta$ and such that

$$A(m_0) = \int_{M \times M} c_0^T(x, y) d\eta.$$

PROOF. By Proposition 1, there exists a compact set $K \in TM$ such that if $\gamma(t) : [0,T] \longrightarrow M$ is a minimizing extremal, then the lifting $(\gamma(t), \dot{\gamma}(t))$ is contained in K for each $t \in [0,T]$. We shall prove that, for each probability measure $\eta \in \mathcal{B}(M \times M)$, there exists a probability measure $m_0 \in \mathcal{B}(K)$ such that $(\pi \times (\pi \circ \psi_0^T))_{\sharp} m_0 = \eta$ and such that

$$A(m_0) = \int_{M \times M} c_0^T(x, y) d\eta.$$

Observing that

- the mappings $m_0 \longrightarrow (\pi \times (\pi \circ \psi_0^T))_{\sharp} m_0$ and $m_0 \longmapsto A(m_0)$ are linear and continuous on the space $\mathcal{B}_1(K)$ of probability measures supported on K,
- the set $\mathcal{B}_1(K)$ is compact for the weak topology, and the action A is continuous on this set,
- the set of probability measures on $M \times M$ is the compact convex closure of the set of Dirac probability measures (probability measures supported in one point), see e. g. [10], p. 73,

it is enough to prove the result when η is a Dirac probability measure (or equivalently when μ_0 and μ_T are Dirac probability measures). Let η be the Dirac probability measure supported at $(x_0, x_1) \in M \times M$. Let $\gamma(t) : [0, T] \longrightarrow M$ be a minimizing extremal with boundary conditions $\gamma(0) = x_0$ and $\gamma(T) = x_1$. In view of the choice of K, we have $(\gamma(0), \dot{\gamma}(0)) \in K$. Let m_0 be the

Dirac probability measure supported at $(\gamma(0), \dot{\gamma}(0))$. It is straightforward that m_t is then the Dirac measure supported at $(\gamma(t), \dot{\gamma}(t))$, so that

$$A(m_0) = \int_0^T L dm_t dt = \int_0^T L(\gamma(t), \dot{\gamma}(t), t) dt = c_0^T(x_0, x_1) = \int_{M \times M} c_0^T d\eta$$

and

$$(\pi \times (\pi \circ \psi_0^T))_{\sharp} m_0 = \eta.$$

Although we are going to build minimizers by other means, we believe the following result is worth being mentioned.

Lemma 7. For each real number a, the set $\mathcal{M}^a(\mu_0, \mu_T)$ of transport measures m which satisfy $A(m) \leq a$, as well as the set $\mathcal{I}^a(\mu_0, \mu_T)$ of initial transport measures m_0 which satisfy $A_0^T(m_0) \leq a$, are compact. As a consequence, there exist optimal initial transport measures, and optimal transport measures.

Proof. This is an easy application of the Prohorov theorem, see the Appendix. $\hfill\Box$

Now we have seen that the problem of finding optimal transport measures is well-posed, let us describe its solutions.

Theorem 1. We have

$$C_0^T(\mu_0, \mu_T) = \min_{m \in \mathcal{M}(\mu_0, \mu_T)} A(m) = \min_{m_0 \in \mathcal{I}(\mu_0, \mu_T)} A(m_0).$$

The mapping

$$m_0 \longmapsto m = ((\psi_0^t)_{\sharp} m_0) \otimes dt$$

between the set \mathcal{OI} of optimal initial measures and the set \mathcal{OM} of optimal transport measures is a bijection. There exists a bounded and locally Lipschitz vectorfield $X(x,t): M \times]0, T[\longrightarrow TM$ such that, for each optimal initial measure $m_0 \in \mathcal{OI}$, the measure $m_t = (\psi_0^t)_{\sharp} m_0$ is supported on the graph of X_t for each $t \in]0,T[$.

The proof will be given in section 4.3. Let us just notice now that the inequalities

$$C_0^T(\mu_0, \mu_T) \geqslant \min_{m_0 \in \mathcal{I}(\mu_0, \mu_T)} A(m_0) \geqslant \min_{m \in \mathcal{M}(\mu_0, \mu_T)} A(m)$$

hold in view of Proposition 6.

2.2 Currents

This formulation finds its roots on one hand in the works of Benamou and Brenier, see [6], and then Brenier, see [13], and on the other hand in the work of Bangert [5]. Let $\Omega^0(M \times [0,T])$ be the set of continuous one-forms on $M \times [0,T]$, endowed with the uniform norm. We will often decompose forms $\omega \in \Omega^0(M \times [0,T])$ as

$$\omega = \omega^x + \omega^t dt,$$

where ω^x is a time-dependent form on M and ω^t is a continuous function on $M \times [0,T]$. To each continuous linear form χ on $\Omega^0(M \times [0,T])$, we associate its time component μ_{χ} , which is the measure on $M \times [0,T]$ defined by

$$\int_{M\times[0,T]}fd\mu_\chi=\chi(fdt)$$

for each continuous function f on $M \times [0,T]$. A Transport current between μ_0 and μ_T is a continuous linear form χ on $\Omega^0(M \times [0,T])$ which satisfies the two conditions:

- 1. The measure μ_{χ} is non negative (and bounded).
- 2. $d\chi = \mu_T \otimes \delta_T \mu_0 \otimes \delta_0$, which means that

$$\chi(df) = \int_{M} f_T d\mu_T - \int_{M} f_0 d\mu_0$$

for each smooth (or equivalently C^1) function $f: M \times [0,T] \longrightarrow \mathbb{R}$.

We call $\mathcal{C}(\mu_0, \mu_T)$ the set of transport currents from μ_0 to μ_T . The set $\mathcal{C}(\mu_0, \mu_T)$ is a closed convex subset of $\left[\Omega^0(M \times [0,T])\right]^*$. We will endow $\mathcal{C}(\mu_0, \mu_T)$ with the weak topology obtained as the restriction of the weak-* topology of $\left[\Omega^0(M \times [0,T])\right]^*$. Transport currents should be thought of as vectorfields whose components are measures, the last component being μ_{χ} .

If Z is a bounded measurable vectorfield on $M \times [0,T]$, and if ν is a finite non-negative measure on $M \times [0,T]$, we define the current $Z \wedge \nu$ by

$$Z \wedge \nu(\omega) := \int_{M \times [0,T]} \omega(Z) d\nu.$$

Every transport current can be written in this way, see [22] or [25]. As a consequence, currents extend as linear forms on the set $\Omega_{\infty}(M \times [0,T])$ of bounded measurable one-forms. If I is a Borel subset of the interval [0,T], it is therefore possible to define the restriction χ_I of the current χ to I by the formula $\chi_I(\omega) = \chi(1_I\omega)$, where 1_I is the indicatrix of I.

Lemma 8. If χ is a transport current, then

$$\tau_{\sharp}\mu_{\chi}=dt,$$

where τ is the projection onto [0,T], see appendix. As a consequence, there exists a measurable familty $\mu_t, t \in]0,T[$ of probability measures on M such that $\mu_{\chi} = \mu_t \otimes dt$, see appendix. There exists a set $I \subset]0,T[$ of total measure such that the relation

$$\int_{M} f_t d\mu_t = \int_{M} f_0 d\mu_0 + \chi_{[0,t[}(df))$$
 (6)

holds for each C^1 function $f: M \times [0,T] \longrightarrow M$ and each $t \in I$.

PROOF. Let $g:[0,T]\longrightarrow \mathbb{R}$ be a continuous function. Setting $G(t)=\int_0^t g(s)ds$, we observe that

$$\int_{M \times [0,T]} g d\mu_{\chi} = \chi(dG) = \int_{M} G_{T} d\mu_{T} - \int_{M} G_{0} d\mu_{0} = G(T) - G(0) = \int_{0}^{T} g(s) ds.$$

This implies that $\tau_{\sharp}\mu_{\chi}=dt$. As a consequence, the measure μ_{χ} can be desintegrated as $\mu_{\chi}=\mu_{t}\otimes dt$. We claim that, for each C^{1} function $f:M\times[0,T]\longrightarrow M$, the relation (6) holds for almost every t. Since the space $C^{1}(M\times[0,T],\mathbb{R})$ is separable, the claim implies the existence of a set $I\subset]0,T[$ of full Lebesgue measure such that (6) holds for all $t\in I$ and all $f\in C^{1}(M\times[0,T],\mathbb{R})$. In order to prove the claim, let us fix a function f in $C^{1}(M\times[0,T],\mathbb{R})$. For each function $g\in C^{1}([0,T],\mathbb{R})$, we have

$$\chi(d(gf)) = \chi(g'fdt) + \chi(gdf)$$

hence

$$g(T) \int_{M} f_{T} d\mu_{T} - g(0) \int_{M} f_{0} d\mu_{0} = \int_{0}^{T} g'(t) \int_{M} f_{t} d\mu_{t} dt + \chi(g df).$$

By applying this relation to a sequence of C^1 functions g approximating $1_{[0,t[}$, we get, at the limit

$$-\int_{M} f_0 d\mu_0 = -\int_{M} f_t d\mu_t + \chi_{[0,t[}(df)$$

at every Lebesgue point of the function $t \to \int_M f_t d\mu_t$.

If $\mu_0 = \mu_T$, an easy example of transport current is given by $\chi(\omega) = \int_M \int_0^T \omega^t dt d\mu_0$. Here are some more interesting examples.

Regular transport currents. The transport current χ is called regular if there exists a bounded measurable section X of the projection $TM \times [0,T] \longrightarrow M \times [0,T]$, and a a nonnegative measure μ on $M \times [0,T]$ such that $\chi = (X,1) \wedge \mu$. The time component of the current $(X,1) \wedge \mu$ is μ . In addition, if $(X,1) \wedge \mu = (X',1) \wedge \mu$ for two vectorfields X and X', then X and X' agree μ -almost everywhere.

The current $\chi = (X, 1) \wedge \mu$, with X bounded, is a regular transport current if and only if there exists a (unique) continuous family $\mu_t \in \mathcal{B}_1(M), t \in [0, T]$ (where μ_0 and μ_T are the transported measures) such that $\mu_{\chi} = \mu_t \otimes dt$ and such that the transport equation

$$\partial_t \mu_t + \partial_x \cdot (X \mu_t) = 0$$

holds in the sense of distributions on $M \times]0,T[$. The relation

$$\int_{M} f_t d\mu_t - \int_{M} f_s d\mu_s = \chi_{[s,t[}(df)$$

then holds for each C^1 function f and each $s \leq t$ in [0,T].

In order to prove that the family μ_t can be chosen continuous, pick a function $f \in C^1(M, \mathbb{R})$ and notice that the equation

$$\int_{M} f d\mu_{t} - \int_{M} f d\mu_{s} = \chi_{[s,t[}(df)) = \int_{s}^{t} \int_{M} df \cdot X_{\sigma} d\mu_{\sigma} d\sigma$$

holds for all $s \leq t$ in a subset of total measure $I \subset [0,T]$. Note that this relation also holds if s = 0 and $t \in I$ and if $s \in I$ and t = T. Since the function $\sigma \longmapsto \int_M df \cdot X_\sigma d\mu_\sigma$ is bounded, we conclude that the function $t \longmapsto \int_M f d\mu_t$ is Lipschitz on $I \cup \{0,T\}$ for each $f \in C^1(M,\mathbb{R})$, with a Lipschitz constant which depends only on $\|df\|_{\infty} \cdot \|X\|_{\infty}$. The family μ_t is then Lipschitz on $I \cup \{0,T\}$ for the 1-Wasserstein distance on probability measures, see [39, 17, 4] for example, the Lipschitz constant depending only on $\|X\|_{\infty}$. It suffices to remember that, on the compact manifold M, the 1-Wasserstein distance on probabilities is topologically equivalent to the weak topology, see for example [41], (48.5) or [39].

Smooth transport currents. A regular transport current is said smooth if it can be written on the form $(X,1) \wedge \lambda$ with a bounded vectorfield X smooth on $M \times]0,T[$ and a measure λ that has a positive smooth density with respect to the Lebesgue class in any chart in $M \times]0,T[$. Every transport current in $\mathcal{C}(\mu_0,\mu_T)$ can be approximated by smooth transport currents, but we shall not use such approximations.

Lipschitz regular transport currents. A regular transport current is said Lipschitz regular if it can be written in the form $(X,1) \wedge \mu$ with a vectorfield X which is bounded and locally Lipschitz on $M \times]0,T[$. Smooth currents are Lipschitz regular. Lipschitz regular transport currents have a remarkable structure:

If $\chi = (X, 1) \wedge \mu$ is a Lipschitz regular transport current with X bounded and locally Lipschitz on $M \times]0, T[$, then we have

$$(\Psi_s^t)_{\sharp}\mu_s = \mu_t$$

where $\Psi_s^t, (s,t) \in]0, T[^2, denotes the flow of the Lipschitz vectorfield X from time s to time t, and <math>\mu_t$ is the continuous family of probability measures such that $\mu_{\chi} = \mu_t \otimes dt$.

This statement follows from standard representation results for solutions of the transport equation, see for example [3] or [4].

Transport current induced from a transport measure. To a transport measure m, we associate the transport current χ_m defined by

$$\chi_m(\omega) = \int_{TM \times [0,T]} \left(\omega^x(x,t) \cdot v + \omega^t(x,t) \right) dm(x,v,t)$$

where the form ω is decomposed as $\omega = \omega^x + \omega^t dt$. Note that the time component of the current χ_m is $\pi_{\sharp}m$. We will see in Lemma 11 that

$$A(\chi_m) \leqslant A(m)$$

with the following definition of the action $A(\chi)$ of a current, with equality if m is concentrated on the graph of any bounded vectorfield $M \times [0,T] \to TM$.

Lemma 9. For each transport current χ , the numbers

$$A_{1}(\chi) = \sup_{\omega \in \Omega^{0}} \left(\chi(\omega^{x}, 0) - \int_{M \times [0, T]} H(x, \omega^{x}(x, t), t) d\mu_{\chi} \right)$$

$$A_{2}(\chi) = \sup_{\omega \in \Omega^{0}} \left(\chi(\omega) - \int_{M \times [0, T]} \left(H(x, \omega^{x}(x, t), t) + \omega^{t} \right) d\mu_{\chi} \right)$$

$$A_{3}(\chi) = \sup_{\omega \in \Omega^{0}} \left(\chi(\omega) - T \sup_{(x, t) \in M \times [0, T]} \left(H(x, \omega^{x}(x, t), t) + \omega^{t} \right) \right)$$

$$A_{4}(\chi) = \sup_{\omega \in \Omega^{0}; \omega^{t} + H(x, \omega^{x}, t) \leq 0} \chi(\omega)$$

$$A_{5}(\chi) = \sup_{\omega \in \Omega^{0}; \omega^{t} + H(x, \omega^{x}, t) \equiv 0} \chi(\omega),$$

are equal. In addition the number $A_i^{\infty}(\chi)$ obtained by replacing in the above suprema the set Ω^0 of continuous forms by the set Ω_{∞} of bounded measurable forms also have the same value.

The last remark in the statement has been added in the last version of the paper and is inspired from [15].

PROOF. It is straightforward that $A_1 = A_2$, this just amounts to simplifying the term $\int \omega^t d\mu_{\chi}$. Since μ_{χ} is a non-negative measure which satisfies $\int_{M \times [0,T]} 1 d\mu_{\chi} = T$, we have

$$\int_{M\times[0,T]} \left(H(x,\omega^x(x,t),t) + \omega^t \right) d\mu_{\chi} \leqslant T \sup_{(x,t)\in M\times[0,T]} \left(H(x,\omega^x(x,t),t) + \omega^t \right)$$

so that $A_3(\chi) \leq A_2(\chi)$. In addition, we obviously have $A_5(\chi) \leq A_4(\chi) \leq A_3(\chi)$. Now notice, in A_2 , that the quantity

$$\chi(\omega) - \int_{M \times [0,T]} \left(H(x, \omega^x(x,t), t) + \omega^t \right) d\mu_{\chi}$$

does not depend on ω^t . Let us consider the form $\tilde{\omega} = (\omega^x, -H(x, \omega^x, t))$, which satisfies the equality $H(x, \tilde{\omega}^x, t) + \tilde{\omega}^t \equiv 0$. We get, for each form ω ,

$$\chi(\omega^x, 0) - \int_{M \times [0,T]} H(x, \omega^x(x,t), t) d\mu_{\chi} = \chi(\tilde{\omega}) \leqslant A_5(\chi)$$

Hence $A_1(\chi) \leq A_5(\chi)$. Exactly the same proof shows that the numbers $A_i^{\infty}(\chi)$ are equal. In order to end the proof, it is enough to check that $A_2(\chi) = A_2^{\infty}(\chi)$. Writing the current χ on the form $Z \wedge \nu$ with a bounded vectorfield Z and a measure $\nu \in \mathcal{B}_+(M \times [0,T])$, we have

$$A_2(\chi) = \sup_{\omega \in \Omega^0} \left(\int_{M \times [0,T]} \omega(Z) d\nu - \int_{M \times [0,T]} \left(H(x, \omega^x(x,t), t) + \omega^t \right) d\mu_\chi \right)$$

and

$$A_2^{\infty}(\chi) = \sup_{\omega \in \Omega_{\infty}} \Big(\int_{M \times [0,T]} \omega(Z) d\nu - \int_{M \times [0,T]} \Big(H(x,\omega^x(x,t),t) + \omega^t \Big) d\mu_{\chi} \Big).$$

The desired result follows by density of continuous functions in $L^1(\nu + \mu_{\chi})$.

Definition 10. We denote by $A(\chi)$ and call action of the transport current χ the common value of the numbers $A_i(\chi)$.

The existence of currents of finite action follows from the following:

Lemma 11. We have

$$A(\chi) = \int_{M \times [0,T]} L(x,X(x,t),t) d\mu$$

for each regular current $\chi = (X,1) \wedge \mu$. If m is a transport measure, and if χ_m is the associated transport current, then $A(\chi_m) \leq A(m)$, with equality if m is supported on the graph of a bounded Borel vectorfield. As a consequence, we have the inequalities

$$C_0^T(\mu_0, \mu_T) \geqslant \min_{m_0 \in \mathcal{I}(\mu_0, \mu_T)} A(m_0) \geqslant \min_{m \in \mathcal{M}(\mu_0, \mu_T)} A(m) \geqslant \min_{\chi \in \mathcal{C}(\mu_0, \mu_T)} A(\chi).$$

PROOF. For each bounded measurable form ω , we have

$$\int_{M\times[0,T]} \omega^x(X) - H(x,\omega^x(x,t),t)d\mu \leqslant \int_{M\times[0,T]} L(x,X(x,t),t)d\mu,$$

so that

$$A((X,1) \wedge \mu) \leqslant \int_{M \times [0,T]} L(x,X(x,t),t) d\mu.$$

On the other hand, taking the form $\omega_0^x(x,t) = \partial_v L(x,X(x,t),t)$ we obtain the pointwise equality

$$L(x, X(x,t), t) = \omega_0^x(X) - H(x, \omega_0^x(x,t), t)$$

and by integration

$$\int_{M \times [0,T]} L(x, X(x,t), t) d\mu = \int_{M \times [0,T]} \omega_0^x(X) - H(x, \omega_0^x(x,t), t) d\mu \leqslant A((X,1) \wedge \mu).$$

This ends the proof of the equality of the two forms of the action of regular currents. Now if χ_m is the current associated to a transport measure m, then we have, for each bounded form $\omega \in \Omega^0(M \times [0,T])$,

$$\chi_m(\omega) - \int_{M \times [0,T]} \omega^t(x,t) + H(x,\omega^x(x,t),t) d\mu_{\chi} = \int_{TM \times [0,T]} \omega^x(v) - H(x,\omega^x(x,t),t) dm$$

by definition of χ_m , so that

$$A(\chi_m) \leqslant \int_{TM \times [0,T]} L(x,v,t) dm = A(m)$$

by the Legendre inequality. In addition, if there exists a bounded measurable vectorfield $X: M \times [0,T] \longrightarrow TM$ such that the graph of $X \times \tau$ supports m, then we can consider the form ω_0^T associated to X as above, and we get the equality for this form.

Although we are going to provide explicitly a minimum of A, we believe the following Lemma is worth being mentioned.

Lemma 12. The functional $A: \mathcal{C}(\mu_0, \mu_T) \longrightarrow \mathbb{R} \cup \{+\infty\}$ is convex and lower semi-continuous, both for the strong and weak-* topologies on $[\Omega^0(M \times [0,T])]^*$. Moreover it is coercive with respect to the strong topology and hence it has a minimum.

PROOF. First note that $A(\chi) < \infty$ if χ is the transport current corresponding to an initial transport measure in $\mathcal{M}(\mu_0, \mu_T)$ arising from a transport plan. Let us define the continuous convex function $\mathbb{H}_T : \Omega^0(M \times [0, T]) \longrightarrow \mathbb{R}$ by

$$\mathbb{H}_T(\omega) = T \sup_{(x,t) \in M \times [0,T]} H(x, \omega^x(x,t), t) + \omega^t.$$

Then the action is the restriction to $\mathcal{C}(\mu_0, \mu_T)$ of the Fenchel conjugate $A = \mathbb{H}^* : [\Omega^0(M \times [0,T])]^* \longrightarrow \mathbb{R} \cup \{+\infty\}$. In other words, A is the supremum over ω of the family of affine functionals

$$\chi \to \chi(\omega) - \mathbb{H}_T(\omega)$$

that are continuous both for the strong and weak- * topologies. Hence A is convex and lower semi-continuous for both topologies. Since

$$A(\chi) \geqslant \sup_{\|\omega\| \leqslant 1} \chi(\omega) - \sup_{\|\omega\| \leqslant 1} \mathbb{H}_T(\omega),$$

A is coercive. The existence of a minimizer is standard: any minimizing sequence (χ_n) is bounded (thanks to coercivity) and has a weakly-* convergent subsequence (because $\Omega^0(M \times [0,T])$ is a separable Banach space). By lower semicontinuity, its weak-* limit is a minimizer. Note that $\mathcal{C}(\mu_0, \mu_T)$ is weakly* closed.

Theorem 2. We have

$$C_0^T(\mu_0, \mu_T) = \min_{\chi \in \mathcal{C}(\mu_0, \mu_T)} A(\chi)$$

where the minimum is taken on all transport currents from μ_0 to μ_T . Every optimal transport current is Lipschitz regular. Let $\chi = (X,1) \wedge \mu$ be an optimal transport current, with X locally Lipschitz on $M \times]0,T[$. The measure $m = (X \times \tau)_{\sharp} \mu \in \mathcal{B}_+(TM \times]0,T[)$ is an optimal transport measure, and χ is the transport current induced from m. Here $\tau : TM \times [0,T] \longrightarrow [0,T]$ is the projection on the second factor, see appendix. We have

$$C_0^T(\mu_0, \mu_T) = A(m) = A(\chi) = \int_{M \times [0,T]} L(x, X(x,t), t) d\mu_{\chi}.$$

This result will be proved in 4.1 after some essential results on the dual approach have been established.

3 Hamilton-Jacobi equation

Most of the results stated so far can be proved by direct approaches using Mather's shortening Lemma, which in a sense is an improvement of the initial observation of Monge, see [33] and [5]. We shall however base our proofs on the use of the Hamilton-Jacobi equation, in the spirit of Fathi's approach to Mather theory, see [20], which should be associated to Kantorovich dual approach of the transportation problem.

3.1 Viscosity solutions and semi-concave functions

It is certainly useful to recall the main properties of viscosity solutions in connection with semiconcave functions. We will not give proofs, and instead refer to [20], [21], [14], as well as the appendix in [8]. We will consider the Hamilton-Jacobi equation

$$\partial_t u + H(x, \partial_x u, t) = 0. (HJ)$$

The function $u: M \times [0,T] \longrightarrow M$ is called K-semi-concave if, for each chart $\theta \in \Theta$ (see appendix), the function

$$(x,t) \longmapsto u(\theta(x),t) - K(||x||^2 + t^2)$$

is concave on $B_3 \times [0,T]$. The function u is called semi-concave if it is K-semi-concave for some K. A function $u: M \times]0, T[\longrightarrow M$ is called locally semi-concave if it is semi-concave on each $M \times [s,t]$, for 0 < s < t < T. The following regularity result follows from Fathi's work, see [20] and also [8].

Proposition 13. Let u_1 and u_2 be two K-semi-concave functions. Let A be the set of minima of the function $u_1 + u_2$. Then the functions u_1 and u_2 are differentiable on A, and $du_1(x,t) + du_2(x,t) = 0$ at each point of $(x,t) \in A$. In addition, the mapping $du_1 : M \times [0,T] \longrightarrow T^*M$ is CK-Lipschitz continuous on A, where C is a universal constant.

Definition 14. We say that the function $u: M \times]s, t[\longrightarrow \mathbb{R}$ is a viscosity solution of (HJ) if the equality

$$u(x,\sigma) = \min_{y \in M} u(y,\zeta) + c_{\zeta}^{\sigma}(y,x)$$

holds for all $x \in M$ and all $s < \zeta < \sigma < t$.

We say that the function $\check{u}: M \times]s, t[\longrightarrow \mathbb{R}$ is a backward viscosity solution of (HJ) if the equality

holds for all $x \in M$ and all $s < \sigma < \zeta < t$.

We say that the function $v: M \times]s, t[\longrightarrow \mathbb{R}$ is a viscosity sub-solution of (HJ) if the inequality

$$v(x,\sigma) \leqslant v(y,\zeta) + c_{\mathcal{E}}^{\sigma}(y,x)$$

holds for all x and y in M and all $s < \zeta < \sigma < t$.

Finally we will say that the function $v: M \times [s,t] \longrightarrow \mathbb{R}$ is a continuous viscosity solution (subsolution, backward solution) of (HJ) if it is continuous on $M \times [s,t]$ and if if $v|_{M \times [s,t[}$ is a viscosity solution of (HJ) (subsolution, backward solution).

Notice that both viscosity solutions and backward viscosity solutions are viscosity subsolutions. That these definitions are equivalent in our setting to the usual ones is studied in the references listed above, but is not useful for our discussion. The only fact which will be used is that, for a C^1 function $u: M \times]s, t[\longrightarrow \mathbb{R}$, being a viscosity solution (or a backward viscosity solution) is equivalent to being a pointwise solution of (HJ), and being a viscosity subsolution is equivalent to satisfy pointwise the inequality $\partial_t u + H(x, \partial_x u, t) \leq 0$.

Differentiability of viscosity solutions. Let $u \in C(M \times [0, T[, \mathbb{R})$ be a viscosity solution of (HJ) (on the interval [0, T[)). We have the expression

$$u(x,t) = \min_{\gamma} u(\gamma(0),0) + \int_{0}^{t} L(\gamma(\sigma),\dot{\gamma}(\sigma),\sigma)d\sigma$$

where the minimum is taken on the set of curves $\gamma \in C^2([s,t],M)$ which satisfy the final condition $\gamma(t) = x$. Let us denote by $\Gamma(x,t)$ the set of minimizing curves in this expression, which are

obviously minimizing extremals of L. We say that $p \in T_x^*M$ is a proximal super-differential of the function $u: M \longrightarrow \mathbb{R}$ at point x if there exists a smooth function $f: M \longrightarrow \mathbb{R}$ such that f - u has a minimum at x and $d_x f = p$.

Proposition 15. Let us fix a point $(x,t) \in M \times]0,T[$. The function u_t is differentiable at x if and only if the set $\Gamma(x,t)$ contains a single element γ , and then $\partial_x u(x,t) = \partial_v L(x,\dot{\gamma}(t),t)$.

For all $(x,t) \in M \times]0,T[$ and $\gamma \in \Gamma(x,t)$, we set $p(s) = \partial_v L(\gamma(s),\dot{\gamma}(s),s)$. Then p(0) is a proximal sub-differential of u_0 at $\gamma(0)$, and p(t) is a proximal super-differential of u_t at x.

We finish with an important property on regularity of viscosity solutions:

Proposition 16. For each continuous function $u_0: M \longrightarrow \mathbb{R}$, the viscosity solution

$$u(x,t) := \min_{y \in M} u_0(y) + c_0^t(y,x)$$

is locally semi-concave on]0,T]. If in addition the initial condition u_0 is Lipschitz, then u is Lipschitz on [0,T].

For each continuous function $u_T: M \longrightarrow \mathbb{R}$, the viscosity solution

$$\breve{u}(x,t) := \max_{y \in M} u_T(y) - c_t^T(x,y)$$

is locally semi-convex on [0,T[. If in addition the final condition u_T is Lipschitz, then u is Lipschitz on [0,T].

PROOF. The part concerning semi-concavity of u is proved in [14], for example. It implies that the function u is locally Lipschitz on]0,T], hence differentiable almost everywhere. In addition, at each point of differentiability of u, we have $\partial_t u + H(x, \partial_x u, t) = 0$ and $\partial_x u(x, t) = p(t) = \partial_v L(x, \dot{\gamma}(t), t)$, where $\gamma : [0, t] \longrightarrow M$ is the only curve in $\Gamma(x, t)$. In order to prove that the function u is Lipschitz, it is enough to prove that there exists a uniform bound on |p(t)|. It is known, see Proposition 15, that $p(0) := \partial_v L(\gamma(0), \dot{\gamma}(0), 0)$ is a proximal sub-differential of the function u_0 at point $\gamma(0)$. If u_0 is Lipschitz, its sub-differentials are bounded: There exists a constant K such that $|p(0)| \leq K$. By completeness, there exists a constant K', which depends only on the Lipschitz constant of u_0 , such that $|p(s)| \leq K'$ for all $s \in [0, t]$. This proves that the function u is Lipschitz. The statements concerning u are proved in a similar way.

3.2 Viscosity solutions and Kantorovich optimal pairs

Given a Kantorovich optimal pair (ϕ_0, ϕ_1) , we define the viscosity solution

$$u(x,t) := \min_{y \in M} \phi_0(x) + c_0^t(y,x)$$

and the backward viscosity solution

which satisfy $u_0 = \check{u}_0 = \phi_0$, and $u_T = \check{u}_T = \phi_1$. Note that both ϕ_1 and $-\phi_0$ are semi-concave hence Lipschitz, that the function u is Lipschitz and locally semi-concave on]0, T], and that the function \check{u} is Lipschitz and locally semi-convex on [0, T].

Proposition 17. We have

$$C_0^T(\mu_0, \mu_T) = \max_{u} \left(\int_M u_T d\mu_T - \int_M u_0 d\mu_0 \right), \tag{7}$$

where the minimum is taken on the set of continuous viscosity solutions $u: M \times [0,T] \longrightarrow \mathbb{R}$ of the Hamilton-Jacobi equation (HJ). The same conclusion holds if the maximum is taken on the set of continuous backward viscosity solutions. The same conclusion also holds if the maximum is taken on the set of continuous viscosity sub-solutions of (HJ).

PROOF. If u(x,t) is a continuous viscosity sub-solution of (HJ), then it satisfies

$$u_T(x) - u_0(y) \leqslant c_0^T(y, x)$$

for each x and $y \in M$, and so, by Kantorovich duality,

$$\left(\int_{M} u_T d\mu_T - \int_{M} u_0 d\mu_0\right) \leqslant C_0^T(\mu_0, \mu_T).$$

The converse inequality is obtained by using the functions u and \check{u} .

Definition 18. If (ϕ_0, ϕ_1) is a Kantorovich optimal pair, then we denote by $\mathcal{F}(\phi_0, \phi_1) \subset C^2([0, T], M)$ the set of curves $\gamma(t)$ such that

$$\phi_1(\gamma(T)) = \phi_0(\gamma(0)) + \int_0^T L(\gamma(t), \dot{\gamma}(t), t) dt.$$

We denote by $\mathcal{T}(\phi_0, \phi_1) \subset M \times]0, T[$ the set

$$\mathcal{T}(\phi_0, \phi_1) = \{ (\gamma(t), t), t \in]0, T[, \gamma \in \mathcal{F}(\phi_0, \phi_1)] \}$$

and by $\tilde{T}(\phi_0, \phi_1) \subset TM \times]0, T[$ the set

$$\tilde{\mathcal{T}}(\phi_0, \phi_1) = \{ (\gamma(t), \dot{\gamma}(t), t), t \in]0, T[, \gamma \in \mathcal{F}(\phi_0, \phi_1)] \},$$

which is obviously invariant under the Euler-Lagrange flow.

Proposition 19. Let (ϕ_0, ϕ_1) be a Kantorovich optimal pair, and let u and \check{u} be the associated viscosity and backward viscosity solutions.

1. We have $\breve{u} \leq u$, and

$$\mathcal{T}(\phi_0, \phi_1) = \{(x, t) \in M \times]0, T[\text{ such that } u(x, t) = \breve{u}(x, t)\}.$$

- 2. At each point $(x,t) \in \mathcal{T}(\phi_0,\phi_1)$, the functions u and \check{u} are differentiable, and satisfy $du(x,t) = d\check{u}(x,t)$. In addition, the mapping $(x,t) \longmapsto du(x,t)$ is locally Lipschitz on $\mathcal{T}(\phi_0,\phi_1)$.
- 3. If $\gamma(t) \in \mathcal{F}(\phi_0, \phi_1)$, then $\partial_x u(\gamma(t), t) = \partial_v L(\gamma(t), \dot{\gamma}(t), t)$. As a consequence, the set

$$\mathcal{T}^*(\phi_0,\phi_1) := \{(x,p,t) \in T^*M \times]0,T[\text{ such that } (x,t) \in \mathcal{T} \text{ and } p = \partial_x u(x,t) = \partial_x \check{u}(x,t) \}$$

is invariant under the Hamiltonian flow, and the restriction to $\tilde{\mathcal{T}}(\phi_0, \phi_1)$ of the projection π is a bi-locally-Lipschitz homeomorphism onto its image $\mathcal{T}(\phi_0, \phi_1)$.

PROOF. Let us fix a point $(x,t) \in M \times]0,T[$. There exist points y and z in M such that $u(x,t) = \phi_0(y) + c_0^t(y,x)$ and $\check{u}(x,t) = \phi_1(z) - c_t^T(x,z)$, so that

$$u(x,t) - \breve{u}(x,t) = \phi_0(y) - \phi_1(z) + c_0^t(y,x) + c_t^T(x,z)$$

$$\geqslant c_0^T(y,z) - (\phi_1(z) - \phi_0(y)) \geqslant 0.$$

In case of equality, we must have $c_0^T(y,z) = c_0^t(y,x) + c_t^T(x,z)$. Let $\gamma_1(s) \in C^2([0,t],M)$ satisfy $\gamma_1(0) = y, \ \gamma_1(t) = x$ and $\int_0^t L(\gamma_1(s),\dot{\gamma}_1(s),s)ds = c_0^t(y,x)$, and let $\gamma_2(s) \in C^2([t,T],M)$ satisfy $\gamma_2(t) = x, \ \gamma_2(T) = z$ and $\int_0^t L(\gamma_2(s),\dot{\gamma}_2(s),s)ds = c_t^T(x,z)$. The curve $\gamma:[0,T] \longrightarrow M$ obtained by pasting γ_1 and γ_2 clearly satisfies $\int_0^T L(\gamma(s),\dot{\gamma}(s),s)ds = c_0^T(y,z)$, it is thus a C^2 minimizer, and belongs to $\mathcal{F}(\phi_0,\phi_1)$. As a consequence, we have $(x,t) \in \mathcal{T}(\phi_0,\phi_1)$.

Lemma 20. If v is a viscosity sub-solution of (HJ) satisfying $v_0 = \phi_0$ and $v_T = \phi_1$, then

PROOF. The inequality $\check{u} \leqslant v \leqslant u$ is easy. For example, for a given point (x,t) there exists y in M such that $u(x,t) = \phi_0(y) + c_0^t(y,x)$, and for this value of y, we have $v(x,t) \leqslant \phi_0(y) + c_0^t(y,x)$, hence $v(x,t) \leqslant u(x,t)$. The proof that $\check{u} \leqslant v$ is similar. In order to prove the second part of the lemma, it is enough to prove that $v(\gamma(t),t) = u(\gamma(t),t)$ for each curve $\gamma \in \mathcal{F}(\phi_0,\phi_1)$. Since v is a sub-solution, we have

$$v(\gamma(T), T) \leq v(\gamma(t), t) + c_t^T(\gamma(t), \gamma(T)).$$

On the other hand, we have

Conversely, we have:

$$v(\gamma(t), t) \leqslant u(\gamma(t), t) \leqslant u(\gamma(0), 0) + c_0^t(\gamma(0), \gamma(t)).$$

As a consequence of all these inequalities, we have

 $\check{u} \leqslant v \leqslant u$. If $(x,t) \in \mathcal{T}(\phi_0,\phi_1)$, then we have v(x,t) = u(x,t).

$$\phi_1(\gamma(T)) = v(\gamma(T), T) \leqslant u(\gamma(0), 0) + c_0^t(\gamma(0), \gamma(t)) + c_t^T(\gamma(t), \gamma(T)) \leqslant \phi_0(\gamma(0)) + c_0^T(\gamma(0), \gamma(T))$$

which is an equality because $\gamma \in \mathcal{F}(\phi_0, \phi_1)$. Hence all the inequalities involved are equalities, and we have $v(\gamma(t), t) = u(\gamma(t), t)$.

The end of the proof of the proposition is straightforward. Point 2 follows from Proposition 13 applied to the locally semi-concave functions u and $-\check{u}$. Point 3 follows from Proposition 15. \square

3.3 Optimal C^1 sub-solution

The following result, on which a large part of the present paper is based, is inspired from [21], but seems new in the present context.

Proposition 21. We have

$$C_0^T(\mu_0, \mu_T) = \max_{v} \left(\int_M v_T d\mu_T - \int_M v_0 d\mu_0 \right),$$

where the maximum is taken on the set of Lipschitz functions $v: M \times [0,T] \longrightarrow \mathbb{R}$ which are C^1 on $M \times]0,T[$ and satisfy the inequality

$$\partial_t v(x,t) + H(x, \partial_x v(x,t), t) \leqslant 0 \tag{8}$$

at each point $(x,t) \in M \times]0,T[.$

PROOF. First, let v(x,t) be a continuous function of $M \times [0,T]$ which is differentiable on $M \times [0,T]$, where it satisfies (8). We then have, for each C^1 curve $\gamma(t):[0,T] \longrightarrow M$, the inequality

$$\int_0^T L(\gamma(t), \dot{\gamma}(t), t) dt \geqslant \int_0^T \partial_x v(\gamma(t), t) \cdot \dot{\gamma}(t) - H(\gamma(t), v(\gamma(t), t), t) dt$$

$$\geqslant \int_0^T \partial_x v(\gamma(t), t) \cdot \dot{\gamma}(t) + \partial_t v(\gamma(t), t) dt = v(\gamma(T), T) - v(\gamma(0), 0).$$

As a consequence, we get $v(y,T) - v(x,0) \leq c_0^T(x,y)$ for each x and y, so that

$$\int v_T d\mu_T - \int v_0 d\mu_0 \leqslant C_0^T(\mu_0, \mu_T).$$

The converse follows directly from the next theorem, which is an analog in our context of the main result of [21]. \Box

Theorem 3. For each Kantorovich optimal pair (ϕ_0, ϕ_1) , there exists a Lipschitz function $v: M \times [0,T] \longrightarrow \mathbb{R}$ which is C^1 on $M \times [0,T]$, which coincides with u on $M \times \{0,T\} \cup \mathcal{T}(\phi_0,\phi_1)$, and which satisfies the inequality (8) strictly at each point of $M \times [0,T] - \mathcal{T}(\phi_0,\phi_1)$.

PROOF. The proof of [21] can't be translated to our context in a straightforward way. Our proof is different, and, we believe, simpler. It is based on:

Proposition 22. There exists a function $V \in C^2(M \times [0,T], \mathbb{R})$ which is null on $\mathcal{T}(\phi_0, \phi_1)$ and which is positive on $M \times [0,T] - \mathcal{T}(\phi_0, \phi_1)$, and such that

$$\phi_1(y) = \min_{\gamma(T)=y} \phi_0(\gamma(0)) + \int_0^T L(\gamma(t), \dot{\gamma}(t), t) - V(\gamma(t), t) dt.$$
 (9)

PROOF. Let us define the norm

$$||u||_2 = \sum_{\theta \in \Theta} ||u \circ \theta||_{C^2(B_1 \times [0,T], \mathbb{R})}$$

of functions $u \in C^2(M \times [0,T], \mathbb{R})$, where Θ is the atlas of M defined in the Appendix. Let us denote by U the open set $M \times]0, T[-\mathcal{T}(\phi_0, \phi_1))$. We need a Lemma.

Lemma 23. Let $U_1 \subset U$ be an open set whose closure \bar{U}_1 is compact and contained in U, and let $\epsilon > 0$ be given. There exists a function $V_1 \in C^2(M \times [0,T],\mathbb{R})$, which is positive on U_1 and null outside of \bar{U}_1 which is such that the equality (9) holds with $V = V_1$, and such that $||V_1||_2 \leq \epsilon$.

PROOF. Let us fix the open set U_1 , the pair (ϕ_0, ϕ_1) and $y \in M$. We claim that the minimum in

$$\min_{\gamma(T)=y} \phi_0(\gamma(0)) + \int_0^T L(\gamma(t), \dot{\gamma}(t), t) - V_1(\gamma(t), t) dt$$

is reached at a path γ the graph of which does not meet U_1 , provided that the function V_1 is supported in \overline{U}_1 and is sufficiently small in the C^0 topology. In order to prove the claim, suppose the contrary. There exists a sequence V_n^1 $(n \in \mathbb{N})$ and a sequence γ_n such that

$$\min_{\gamma(T)=y} \phi_0(\gamma(0)) + \int_0^T L(\gamma(t), \dot{\gamma}(t), t) - V_n(\gamma(t), t) dt$$

is reached at γ_n , the graph of γ_n meets U_1 , V_n is supported in \overline{U}_1 (for all $n \in \mathbb{N}$) and $V_n \to 0$ in the C^0 topology. As a consequence each γ_n is C^2 and the sequence γ_n $(n \in \mathbb{N})$ is a minimizing sequence for

$$\phi_1(y) = \min_{\gamma(T) = y} \phi_0(\gamma(0)) + \int_0^T L(\gamma(t), \dot{\gamma}(t), t) dt.$$
 (10)

Hence this sequence is compact for the C^2 topology and, extracting a subsequence if needed, can be assumed to converge to some γ_{∞} . Clearly γ_{∞} is a minimizer for (10) the graph of which

meets \overline{U}_1 . This is a contradiction with $\overline{U}_1 \subset U = M \times]0, T[\setminus \mathcal{T}(\phi_0, \phi_1)]$ and the fact that the graph of γ_{∞} is included in $\overline{\mathcal{T}(\phi_0, \phi_1)}$ (see Definition 18).

Let $U_n \subset U, n \in \mathbb{N}$ be a countable sequence of open sets covering U and whose closures \overline{U}_n are contained in U. There exists a sequence V_n of functions of $C^2(M \times [0,T],\mathbb{R})$ such that, for each $n \in \mathbb{N}$:

- The function V_n is positive in U_n and null outside of \bar{U}_n .
- We have $||V_n||_2 \leqslant 2^{-n}\epsilon$.
- The equality (9) holds for the function $V^n = \sum_{i=1}^n V_i$.

Such a sequence can be build inductively by applying the lemma to the Lagrangian $L - V^{n-1}$ with $\epsilon_n = 2^{-n}\epsilon$. Since $||V_n|| \leq 2^{-n}\epsilon$, the sequence V^n is converging in C^2 norm to a limit $V \in C^2(M \times [0,T],\mathbb{R})$. This function V satisfies the desired properties. The proposition is proved.

In order to finish the proof of the theorem, we shall consider the new Lagrangian $\tilde{L} = L - V$, and the associated Hamiltonian $\tilde{H} = H + V$, as well as the associated cost functions \tilde{c}_s^t . Let

$$\tilde{u}(x,t) := \min_{y \in M} \phi_0(y) + \tilde{c}_0^t(y,x),$$

be the viscosity solution of the Hamilton-Jacobi equation

$$\partial_t \tilde{u} + H(x, \partial_x \tilde{u}, t) = -V(x, t) \tag{HJ}$$

emanating from ϕ_0 . The equality (9) says that $\tilde{u}_T = \phi_1 = u_T$. The function \tilde{u} is Lipschitz on $M \times [0,T]$, as a viscosity solution of $(\tilde{H}J)$ emanating from a Lipschitz function. It is obviously a viscosity sub-solution of the equation (HJ), which is strict outside of $M \times \{0,T\} \cup \mathcal{T}(\phi_0,\phi_1)$ (where V is positive). This means that the inequality (8) is strict at each point of differentiability of \tilde{u} outside of $M \times \{0,T\} \cup \mathcal{T}(\phi_0,\phi_1)$. We have $\tilde{u} \leq \tilde{u} \leq u$, this relation being satisfied by each viscosity sub-solution of (HJ) which satisfies $u_0 = \phi_0$ and $u_T = \phi_1$. As a consequence, we have $\tilde{u} = \tilde{u} = u$ on $\mathcal{T}(\phi_0,\phi_1)$, and the function \tilde{u} is differentiable at each point of $\mathcal{T}(\phi_0,\phi_1)$. Furthermore, we have $du = d\tilde{u} = d\tilde{u}$ on this set. We then obtain the desired function v of the theorem from the function \tilde{u} by regularisation, applying Theorem 9.2 of [21].

4 Optimal objects of the direct problems

We prove Theorem A as well as the results of section 2. The following lemma generalizes a result of Benamou and Brenier, see [6].

Lemma 24. We have the equality

$$C_0^T(\mu_0, \mu_T) = \min_{m_0 \in \mathcal{I}(\mu_0, \mu_T)} A(m_0) = \min_{m \in \mathcal{M}(\mu_0, \mu_T)} A(m) = \min_{\chi \in \mathcal{C}(\mu_0, \mu_T)} A(\chi).$$

Moreover $\chi(dv) = A(\chi)$ for every optimal χ , where v is given by Theorem 3

PROOF. In view of Lemma 11, it is enough to prove that, for each transport current $\chi \in \mathcal{C}(\mu_0, \mu_T)$, we have $A(\chi) \geqslant C_0^T(\mu_0, \mu_T)$. Let $v: M \times [0, T] \longrightarrow \mathbb{R}$ be a Lipschitz sub-solution of (HJ) which is C^1 on $M \times]0, T[$, and such that (v_0, v_T) is a Kantorovich optimal pair. For each current $\chi \in \mathcal{C}(\mu_0, \mu_T)$, we have $A(\chi) \geqslant \chi(dv) = C_0^T(\mu_0, \mu_T)$, which ends the proof.

From now on we choose and fix:

- An optimal Kantorovich pair (ϕ_0, ϕ_1) .
- A Lipschitz sub-solution $v: M \times [0,T] \longrightarrow \mathbb{R}$ of the Hamilton-Jacobi equation which satisfies $v_0 = \phi_0$ and $v_T = \phi_1$ and which is C^1 on $M \times [0,T]$.
- A bounded vectorfield $X(x,t): M \times]0, T[\longrightarrow TM$ which is locally Lipschitz and satisfies

$$X(x,t) = \partial_{p}H(x, \partial_{x}v(x,t), t) \quad \text{on} \quad \mathcal{T}(\phi_{0}, \phi_{1}). \tag{11}$$

4.1 Characterization of optimal currents.

Each optimal transport current χ can be written

$$\chi = (X,1) \wedge \mu_{\chi},$$

with a measure μ_{χ} concentrated on $\mathcal{T}(\phi_0, \phi_1)$. The current χ is then Lipschitz regular, so that there exists a transport interpolation $\mu_t, t \in [0,T]$ such that $\mu_{\chi} = \mu_t \otimes dt$ (see Appendix) and such that $\mu_t = (\Psi_s^t)_{\sharp} \mu_s$ for each s and t in]0,T[.

PROOF. Let χ be an optimal transport current, that is a transport current $\chi \in \mathcal{C}(\mu_0, \mu_T)$ such that $A(\chi) = C_0^T(\mu_0, \mu_T)$. Let us recall the definition of the action $A(\chi)$ that will be used here:

$$A(\chi) = \sup_{\omega \in \Omega^0} \left(\chi(\omega^x, 0) - \int_{M \times [0, T]} H(x, \omega^x(x, t), t) d\mu_{\chi} \right).$$

Since $H(x, \partial_x v, t) + \partial_t v \leq 0$, we have

$$A(\chi) = \chi(dv) \leqslant \chi(dv) - \int H(x, \partial_x v(x, t), t) + \partial_t v d\mu_{\chi} = \chi(\partial_x v, 0) - \int H(x, \partial_x v(x, t), t) d\mu_{\chi}.$$

The other inequality holds by the definition of A, so that

$$\chi(dv) = \chi(dv) - \int H(x, \partial_x v(x, t), t) + \partial_t v d\mu_{\chi} = \chi(\partial_x v, 0) - \int H(x, \partial_x v(x, t), t) d\mu_{\chi},$$

and we conclude that the function $H(x, \partial_x v(x, t), t) + \partial_t v$ vanishes on the support of μ_{χ} , or in other words that the measure μ_{χ} is concentrated on the set $\mathcal{T}(\phi_0, \phi_1)$. In addition, for all form $\omega = \omega^x + \omega^t dt$, we have

$$\chi(\partial_x v + \omega^x, 0) - \int H(x, \partial_x v + \omega^x, t) d\mu_{\chi} \leqslant \chi(\partial_x v, 0) - \int H(x, \partial_x v, t) d\mu_{\chi} = A(\chi).$$

Hence the equality

$$\chi(\omega^x, 0) = \int \partial_p H(x, \partial_x v, t)(\omega^x) d\mu_{\chi}$$

holds for each form ω . This equality can be rewritten

$$\chi(\omega) = \int \partial_p H(x, \partial_x v, t)(\omega^x) + \omega^t d\mu_{\chi}$$

which is precisely saying that

$$\chi = (\partial_n H(x, \partial_x v(x, t), t), 1) \wedge \mu_{\gamma} = (X, 1) \wedge \mu_{\gamma}.$$

The last equality follows from the fact that the vectorfields X and $\partial_p H(x, \partial_x v(x, t), t)$ are equal on the support of μ_χ . By the structure of Lipschitz regular transport currents, we obtain the existence of a continuous family $\mu_t, t \in [0, T]$ of probability measures such that $\mu_\chi = \mu_t \otimes dt$ and such that $\mu_t = (\Psi_s^t)_{\sharp} \mu_s$ for each s and t in]0, T[. Since the restriction to a subinterval $[s, t] \subset [0, T]$ of an optimal transport current χ is clearly an optimal transport current for the transportation problem between μ_s and μ_t with cost c_s^t , we obtain that the path μ_t is a transport interpolation.

4.2 Characterization of transport interpolations.

Each transport interpolation μ_t satisfies

$$\mu_t = (\Psi_s^t)_{\sharp} \mu_s$$

for each $(s,t) \in]0,T[^2]$. The mapping

$$\mu_t \longmapsto (X,1) \wedge (\mu_t \otimes dt)$$

is a bijection between the set of transport interpolations and the set of optimal transport currents.

PROOF. We fix a transport interpolation μ_t and two times s < s' in]0, T[. Let χ_1 be a transport current on $M \times [0, s]$ between the measures μ_0 and μ_s which is optimal for the cost c_0^s , let χ_2 be a transport current on $M \times [s, s']$ between the measures μ_s and $\mu_{s'}$ which is optimal for the cost $c_s^{s'}$ and let χ_3 be a transport current on $M \times [s', T]$ between the measures $\mu_{s'}$ and μ_T which is optimal for the cost $c_{s'}^T$. Then the current χ on $M \times [0, T]$ which coincides with χ_1 on $M \times [0, s]$, with χ_2 on $M \times [s, s']$ and with χ_3 on [s', T] belongs to $\mathcal{C}(\mu_0, \mu_T)$. In addition, since μ_t is a transport interpolation, we have

$$A(\chi) = C_0^s(\mu_0, \mu_s) + C_s^{s'}(\mu_s, \mu_{s'}) + C_{s'}^T(\mu_{s'}, \mu_T) = C_0^T(\mu_0, \mu_T).$$

Hence χ is an optimal transport current for the cost c_0^T . In view of the characterisation of optimal currents, this implies that $\chi = (X, 1) \wedge \mu_{\chi}$, and that

$$\mu_{\chi} = ((\Psi_s^t)_{\sharp} \mu_s) \otimes dt = ((\Psi_{s'}^t)_{\sharp} \mu_{s'}) \otimes dt.$$

By uniqueness of the continuous desintegration of μ_{χ} , we obtain that, for each $t \in]0, T[$, $(\Psi_s^t)_{\sharp}\mu_s = (\Psi_{s'}^t)_{\sharp}\mu_{s'}$, and since this holds for all s and s', that $(\Psi_s^t)_{\sharp}\mu_s = \mu_t$ for all $(s,t) \in]0, T[^2]$. It follows that $\chi = (X,1) \wedge (\mu_t \otimes dt)$. We have proved that the mapping

$$\mu_t \longmapsto (X,1) \land (\mu_t \otimes dt)$$

associates an optimal transport current to each transport interpolation. This mapping is obviously injective, and it is surjective in view of the characterization of optimal currents. \Box

4.3 Characterization of optimal measures.

The mapping

$$\chi \longmapsto (X \times \tau)_{\mathsf{H}} \mu_{\mathsf{Y}}$$

is a bijection between the set of optimal transport currents and the set of optimal transport measures $(\tau: M \times [0,T] \to [0,T]$ is the projection on the second factor; see Appendix). Each optimal transport measure is thus invariant (see (4) and Definition 5). The mapping

$$m_0 \longmapsto \mu_t = (\pi \circ \psi_0^t)_{\sharp} m_0$$

is a bijection between the set of optimal initial measures m_0 and the set of interpolations. An invariant measure m is optimal if and only if it is supported on the set $\tilde{\mathcal{T}}(\phi_0, \phi_1)$.

PROOF. If m is an optimal transport measure, then the associated current χ_m is an optimal transport current, and $A(m) = A(\chi_m)$. Let μ_m be the time component of χ_m , which is also the

measure $(\pi \times \tau)_{\sharp}m$. In view of the characterization of optimal currents, we have $\chi_m = (X, 1) \wedge \mu_m$. We claim that the equality $A(\chi_m) = A(m)$ implies that m is supported on the graph of X. Indeed, we have the pointwise inequality

$$\partial_x v(x,t) \cdot V - H(x, \partial_x v(x,t), t) \leqslant L(x, V, t) \tag{12}$$

for each $(x, V, t) \in TM \times]0, T[$. Integrating with respect to m, we get the equality

$$A(\chi_m) = \chi_m(dv) = \int_{TM \times [0,T]} \partial_x v(x,t) \cdot V + \partial_t v(x,t) dm(x,V,t)$$

$$= \int_{TM \times [0,T]} \partial_x v(x,t) \cdot V - H(x,\partial_x v(x,t),t) dm(x,V,t) = \int_{M \times [0,T]} L(x,V,t) dm(x,V,t) = A(m),$$

which means that m is concentrated on the set where the inequality (12) is an equality, that is on the graph of the vectorfield $\partial_p H(x, \partial_x v(x,t), t)$. Since μ_m is supported on \mathcal{T} , the measure m is supported on $\tilde{\mathcal{T}}$ and satisfies $m = (X \times \tau)_{\sharp} \mu_m$. Let μ_t be the transport interpolation such that $\mu_m = \mu_t \otimes dt$. Setting $m_t = (X_t)_{\sharp} \mu_t$, we have $m = m_t \otimes dt$. Observing that the relation

$$X_t \circ \Psi_s^t = \psi_s^t \circ X_s$$

holds on \mathcal{T}_s , we conclude, since μ_s is supported on \mathcal{T}_s , that

$$(\psi_s^t)_{\sharp} m_s = m_t,$$

which means that the measure m is invariant.

Conversely, let $m = m_t \otimes dt$ be an invariant measure supported on $\tilde{T}(\phi_0, \phi_1)$. We have

$$A(m) = \int_0^T \int_{TM} L(x, v, t) dm_t(x, v) dt = \int_0^T \int_{TM} L((\psi_0^t(x, v), t) dm_0(x, v) dt,$$

and by Fubini,

$$A(m) = \int_{TM} \int_0^T L((\psi_0^t(x, v), t) dt dm_0(x, v)) = \int_{TM} \phi_1(\pi \circ \psi_0^T(x, v)) - \phi_0(x) dm_0(x, v),$$

and since m_0 is an initial transport measure, we get

$$A(m) = \int_{TM} \phi_1 d\mu_T - \int_{TM} \phi_0 d\mu_0 = C_0^T(\mu_0, \mu_T).$$

5 Absolute continuity

In this section, we make the additional assumption that the initial measure μ_0 is absolutely continuous, and prove Theorem B. The following lemma answers a question asked to us by Cedric Villani.

Lemma 25. If μ_0 or μ_T is absolutely continuous with respect to the Lebesgue class, then each interpolating measure $\mu_t, t \in]0, T[$, is absolutely continuous.

PROOF. If $\mu_t, t \in [0,T]$ is a transport interpolation, we have proved that

$$\mu_t = (\pi \circ \psi_s^t \circ X_s)_{\dagger} \mu_s$$

for each $s \in]0,T[$, and $t \in [0,T]$. Since the function $\pi \circ \psi_t^s \circ X_t$ is Lipschitz, it maps Lebesgue zero measure sets into Lebesgue zero measure sets, and so it transport singular measures into singular measures. It follows that if, for some $s \in]0,T[$, the measure μ_s is not absolutely continuous, then none of the measures $\mu_t,t \in [0,T]$ are absolutely continuous.

In order to continue the investigation of the specific properties satisfied when μ_0 is absolutely continuous, we first need some more general results. Let (ϕ_0, ϕ_1) be an optimal Kantorovich pair for the measures μ_0 and μ_T and for the cost c_0^T . Recall that we have defined $\mathcal{F}(\phi_0, \phi_1) \subset C^2([0, T], M)$ as the the set of curves $\gamma(t)$ such that

$$\phi_1(\gamma(T)) = \phi_0(\gamma(0)) + \int_0^T L(\gamma(t), \dot{\gamma}(t), t) dt.$$

Let $\mathcal{F}_0(\phi_0, \phi_1)$ be the set of initial velocities $(x, v) \in TM$ such that the curve $t \longmapsto \pi \circ \psi_0^t(x, v)$ belongs to $\mathcal{F}(\phi_0, \phi_1)$. Note that there is a natural bijection between $\mathcal{F}_0(\phi_0, \phi_1)$ and $\mathcal{F}(\phi_0, \phi_1)$.

Lemma 26. The set $\mathcal{F}_0(\phi_0, \phi_1)$ is compact. The maps π and $\pi \circ \psi_0^T : \mathcal{F}_0(\phi_0, \phi_1) \longrightarrow M$ are surjective. If x is a point of differentiability of ϕ_0 , then the set $\pi^{-1}(x) \cap \mathcal{F}_0(\phi_0, \phi_1)$ contains one and only one point. There exists a Borel measurable set $\Sigma \subset M$ of full measure, whose points are points of differentiability of ϕ_0 , and such that the map

$$x \longmapsto S(x) = \pi^{-1}(x) \cap \mathcal{F}_0(\phi_0, \phi_1)$$

is Borel measurable on Σ .

PROOF. The compactness of $\mathcal{F}_0(\phi_0, \phi_1)$ follows from the fact, already mentioned, that the set of minimizing extremals $\gamma: [0, T] \longrightarrow M$ is compact for the C^2 - topology.

It is equivalent to say that the projection π restricted to $\mathcal{F}_0(\phi_0, \phi_1)$ is surjective, and to say that, for each point $x \in M$, there exists a curve emanating from x in $\mathcal{F}(\phi_0, \phi_1)$. In order to build such curves, recall that

$$\phi_0(x) = \max_{\gamma} \phi_1(\gamma(T)) - \int_0^T L(\gamma(t, \dot{\gamma}(t), t))dt$$

where the maximum is taken on the set of curves which satisfy $\gamma(0) = x$. Any maximizing curve is then a curve of $\mathcal{F}(\phi_0, \phi_1)$ which satisfies $\gamma(0) = x$. In order to prove that the map $\pi \circ \psi_0^T$ restricted to $\mathcal{F}_0(\phi_0, \phi_1)$ is surjective, it is sufficient to build, for each point $x \in M$, a curve in $\mathcal{F}(\phi_0, \phi_1)$ which ends at x. Such a curve is obtained as a minimizer in the expression

$$\phi_1(x) = \min_{\gamma} \phi_0(\gamma(0)) + \int_0^T L(\gamma(t, \dot{\gamma}(t), t))dt.$$

Now let us consider a point x of differentiability of ϕ_0 . Applying the general result on the differentiability of viscosity solutions to the Backward viscosity solution \check{u} , we get that there exists a unique maximizer to the problem

$$\phi_0(x) = \max_{\gamma} \phi_1(\gamma(T)) - \int_0^T L(\gamma(t, \dot{\gamma}(t), t))dt$$

and that this maximizer is the extremal with initial condition $(x, \partial_p H(x, d\phi_0(x), 0))$. As a consequence, there exists one and only one point S(x) in $\mathcal{F}_0(\phi_0, \phi_1)$ above x, and in addition we have the explicit expression

$$S(x) = \partial_p H(x, d\phi_0(x), 0).$$

Since the set of points of differentiability of ϕ_0 has total Lebesgue measure –because ϕ_0 is Lipschitz– there exists a sequence K_n of compact sets such that ϕ_0 is differentiable at each point of K_n and such that the Lebesgue measure of $M-K_n$ is converging to zero. For each n, the set $\pi^{-1}(K_n) \cap \mathcal{F}_0(\phi_0, \phi_1)$ is compact, and the restriction to this set of the canonical projection π is injective and continuous. It follows that the inverse function S is continuous on K_n . As a consequence, the map S is Borel measurable on $\Sigma := \bigcup_n K_n$.

Lemma 27. The initial transport measure m_0 is optimal if and only if it is an initial transport measure supported on $\mathcal{F}_0(\phi_0, \phi_1)$.

PROOF. This statement is a reformulation of the result in 4.3 stating that the optimal transport measures are the invariant measures supported on $\tilde{\mathcal{T}}(\phi_0, \phi_1)$.

Proposition 28. If μ_0 is absolutely continuous, then there exists a unique optimal initial measure m_0 . There exists a Borel section $S: M \longrightarrow TM$ of the canonical projection such that $m_0 = S_{\sharp}\mu_0$, this section is unique μ_0 -almost everywhere. For each $t \in [0,T]$, the map $\pi \circ \psi_0^t \circ S: M \longrightarrow M$ is then an optimal transport map between μ_0 and μ_t .

PROOF. Let $S: \Sigma \longrightarrow TM$ be the Borel map constructed in Lemma 26. For convenience, we shall also denote by S the same map extended by zero outside of Σ , which is a Borel section $S: M \longrightarrow TM$. Since the set Σ is of full Lebesgue measure, and since the measure μ_0 is absolutely continuous, we have $\mu_0(\Sigma) = 1$. Let us consider the measure $m_0 = S_{\sharp}(\mu_{0|\Sigma})$. This is a probability measure on TM, which is concentrated on $\mathcal{F}_0(\phi_0,\phi_1)$, and which satisfies $\pi_{\sharp}m_0 = \mu_0$. We claim that it is the only measure with these properties. Indeed, if \tilde{m}_0 is a measure with these properties, then $\pi_{\sharp}\tilde{m}_0 = \mu_0$, hence the measure \tilde{m}_0 is concentrated on $\pi^{-1}(\Sigma) \cap \mathcal{F}_0(\phi_0,\phi_1)$. But then, since π induces a Borel isomorphism from $\pi^{-1}(\Sigma) \cap \mathcal{F}_0(\phi_0,\phi_1)$ onto its image Σ , of inverse S, we must have $\tilde{m}_0 = S_{\sharp}\mu_0$. As a consequence, the measure $m_0 = S_{\sharp}\mu_0$ is the only candidate to be an optimal initial transport measure. Since we have already proved the existence of an optimal initial transport measure, it implies that m_0 is the only optimal initial transport measure. Of course, we could prove directly that m_0 is an initial transport measure, but as we have seen, it is not necessary.

5.1 Remark

That there exists an optimal transport map if μ_0 is absolutely continuous could be proved directly as a consequence of the following properties of the cost function.

Lemma 29. The cost function $c_0^T(x,y)$ is semi-concave on $M \times M$. In addition, we have the following injectivity property for each $x \in M$: If the differentials $\partial_x c_0^T(x,y)$ and $\partial_x c_0^T(x,y')$ exist and are equal, then y = y'.

In view of these properties of the cost function, it is not hard to prove the following lemma using a Kantorovich optimal pair in the spirit of works of Brenier [12] and Carlier [16].

Lemma 30. There exists a compact subset $K \in M \times M$, such that the fiber $K_x = K \cap \pi_0^{-1}(x)$ contains one and only one point for Lebesgue almost every x, and which contains the support of all optimal plans.

The proof of the existence of an optimal map for an absolutely continuous measure μ_0 can then be terminated using the following result, see [1], Proposition 2.1.

Proposition 31. A transport plan η is induced from a transport map if and only if it is concentrated on a η -measurable graph.

5.2 Remark

Assuming only that μ_0 vanishes on countably (d-1)-rectifiable sets, we can conclude that the same property holds for all interpolating measures $\mu_t, t < T$, and that Proposition 28 hold. This is proved almost identically. The only refinement needed is that the set of singular points of the semi-convex function ϕ_0 is a countably (d-1)-rectifiable, see [14].

6 Aubry-Mather theory

We explain the relations between the results obtained so far and Mather theory, and prove Theorem C. Up to now, we have worked with fixed measures μ_0 and μ_T . Let us study the optimal value $C_0^T(\mu_0, \mu_T)$ as a function of the measures μ_0 and μ_T .

Lemma 32. The function

$$(\mu_0, \mu_T) \longmapsto C_0^T(\mu_0, \mu_T)$$

is convex and lower semi-continuous on the set of pairs of probability measures on M.

PROOF. It follows directly from the expression

$$C_0^T(\mu_0, \mu_T) = \max_{(\phi_0, \phi_1)} \int_M \phi_1 d\mu_T - \int_M \phi_0 d\mu_0$$

as a maximum of continuous linear functions.

From now on, we consider that the Lagrangian L is defined for all times, $L \in C^2(TM \times \mathbb{R}, \mathbb{R})$, and satisfies

$$L(x, v, t + 1) = L(x, v, t)$$

in addition to the standing hypotheses. Let us restate Theorem C with more details. Recall that α is the action of Mather measures, as defined in the introduction.

Theorem C'. There exists a Lipschitz vectorfield X_0 on M such that all the Mather measures are supported on the graph of X_0 . We have

$$\alpha = \min_{\mu} C_0^1(\mu, \mu),$$

where the minimum is taken on the set of probability measures on M. The mapping $m_0 \mapsto (\pi)_{\sharp} m_0$ is a bijection between the set of Mather measures m_0 and the set of probability measures μ on M satisfying $C_0^1(\mu,\mu) = \alpha$. More precisely, if μ is such a probability measure, then there exists one and only one initial transport measure m_0 for the transport problem between $\mu_0 = \mu$ and $\mu_1 = \mu$ with cost c_0^1 , this measure is $m_0 = (X_0)_{\sharp} \mu$, and it is a Mather measure.

The proof, and related digressions, occupy the end of the section.

Lemma 33. The following minima

$$\alpha_T := \min_{\mu \in \mathcal{B}_1(M)} \frac{1}{T} C_0^T(\mu, \mu), T \in \mathbb{N}$$

exist and are all equal. In addition, any measure $\mu^1 \in \mathcal{B}_1(M)$ which is minimizing $C_0^1(\mu, \mu)$ is also minimizing $C_0^T(\mu, \mu)$ for all $T \in \mathbb{N}$.

PROOF. The existence of the minima follows from the compactness of the set of probability measures and from the semi-continuity of the function C_0^T . Let μ^1 be a minimizing measure for α_1 and let m^1 be an optimal transport measure for the transportation problem $C_0^1(\mu^1, \mu^1)$. Let m^T be the measure on $TM \times [0,T]$ obtained by concatenating T translated versions of m^1 . It means that m^T is the only measure on $TM \times [0,T]$ whose restriction to $TM \times [i,i+1]$ is obtained by translation from m, for each integer i. It is easy to check that m^T is indeed a transport measure between the $\mu_0 = \mu^1$ and $\mu_T = \mu^1$ on the times interval [0,T], and that $A_0^T(m^T) = TA_0^1(m^1)$. As a consequence, we have

$$T\alpha_T \leqslant C_0^T(\mu^1, \mu^1) \leqslant A_0^T(m^T) = TC_0^1(\mu^1, \mu^1) = T\alpha_1,$$

which implies the inequality $\alpha_T \leq \alpha_1$.

Let us now prove that $\alpha_T \geqslant \alpha_1$. In order to do so, we consider an optimal measure μ^T for α_T , and consider a transport interpolation μ_t^T , $t \in [0,T]$ between the measures $\mu_0 = \mu^T$ and $\mu_T = \mu^T$. Let us then consider, for $t \in [0,1]$, the measure

$$\tilde{\mu}_t^T := \frac{1}{T} \sum_{i=0}^{T-1} \mu_{t+i}^T,$$

and note that $T\tilde{\mu}_0^T = \mu_0^T + \sum_{i=1}^{T-1} \mu_i^T = \mu_T^T + \sum_{i=1}^{T-1} \mu_i^T = T\tilde{\mu}_1^T$. In view of the convexity of the function C_0^1

$$C_0^1(\tilde{\mu}_0^T, \tilde{\mu}_1^T) = C_0^1 \left(\frac{1}{T} \sum_{i=0}^{T-1} (\mu_i^T, \mu_{i+1}^T) \right) \leqslant \frac{1}{T} \sum_{i=0}^{T-1} C_i^{i+1}(\mu_i^T, \mu_{i+1}^T) = \frac{1}{T} C_0^T(\mu^T, \mu^T) = \alpha_T.$$

Since $\tilde{\mu}_0^T = \tilde{\mu}_1^T$, this implies that $\alpha_1 \leq \alpha_T$, as desired.

Lemma 34. We have $\alpha_1 \leq \alpha$.

PROOF. If m_0 is a Mather measure, then it is an initial measure for the transport problem between $\mu_0 = (\pi)_{\sharp} m_0$ and $\mu_1 = (\pi)_{\sharp} m_0$ for the cost c_0^1 . As a consequence, we have $\alpha = A_0^1(m_0) \geqslant C_0^1(\mu_0, \mu_0) \geqslant \alpha_1$.

Lemma 35. Let μ^1 be a probability measure on M such that $C_0^1(\mu^1, \mu^1) = \alpha_1$. Then there exists a unique initial transport measure m_0 for the transportation problem between $\mu_0 = \mu^1$ and $\mu_1 = \mu^1$ for the cost c_0^1 . This measure satisfies $(\psi_0^1)_{\sharp} m_0 = m_0$. We have $\alpha_1 = A_0^1(m_0) \geqslant \alpha$, so that $\alpha = \alpha_1$ and m_0 is a Mather measure. There exists a constant K, which depends only on L, such that the measure m_0 is supported on the graph of a K-Lipschitz vectorfield.

PROOF. Let us fix a probability measure μ^1 on M such that $C_0^1(\mu^1, \mu^1) = \alpha_1$. Let $X : M \times [0,2] \longrightarrow TM$ be a vectorfield associated to the transport problem $C_0^2(\mu^1, \mu^1)$ by Theorem A. Note that X_1 is Lipschitz on M with a Lipschitz constant K which does not depend on μ_1 . We choose X once and for all and then fix it.

To each optimal transport measure m^1 for the transport problem $C_0^1(\mu^1, \mu^1)$, we associate the transport measure m^2 on $TM \times [0, 2]$ obtained by concatenation of two translated versions of m^1 , as in the proof of Lemma 33. We have

$$A_0^2(m^2) = 2A_0^1(m^1) = 2\alpha_1 = 2\alpha_2 = C_0^2(\mu^1, \mu^1).$$

The measure m^2 is thus an optimal transport measure for the transportation problem $C_0^2(\mu^1, \mu^1)$. Let $m_t, t \in [0, 2]$ be the continuous family of probability measures on TM such that $m^2 = m_t \otimes dt$. Note that $m_t = (\psi_s^t)_{\sharp} m_s$ for all s and t in [0, 2], and that m_0 is the initial transport measure for the transportation problem $C_0^1(\mu^1, \mu^1)$ associated to m^1 . Since the measure m^2 was obtained by concatenation of two translated versions of the same measure m^1 , we must have $m_{t+1} = m_t$ for almost all $t \in]0, 1[$, and, by continuity, $m_0 = m_1 = m_2$. This implies that $m_0 = (\psi_0^1)_{\sharp} m_0$. Finally, the characterization of optimal measures implies that $m_0 = m_1 = (X_1)_{\sharp} \mu^1$. We have proved that the measure $(X_1)_{\sharp} \mu^1$ is the only optimal initial transport measure for the transportation problem $C_0^1(\mu^1, \mu^1)$.

PROOF OF THE THEOREM. Let m_0 be a Mather measure, and let $\mu_0 = \pi_{\sharp} m_0$. Note that we also have $\mu_0 = (\pi \circ \psi_0^1)_{\sharp} m_0$. As a consequence, m_0 is an initial transport measure for the transport between μ_0 and μ_0 for the cost c_0^1 , and we have

$$\alpha = A_0^1(m_0) \geqslant C_0^1(\mu_0, \mu_0) \geqslant \alpha_1.$$

Since $\alpha_1 = \alpha$, all these inequalities are equalities, so that m_0 is an optimal initial transport, and $C_0^1(\mu_0, \mu_0) = \alpha_1$. It follows from Lemma 35 that m_0 is supported on the graph of a K-Lipschitz vectorfield.

Up to now, we have proved that each Mather measure is supported on the graph of a K-Lipschitz vectorfield. There remains to prove that all Mather measures are supported on a single K-Lipschitz graph. In order to prove this, let us denote by $\tilde{\mathcal{M}} \subset TM$ the union of the supports of Mather measures. If (x,v) and (x',v') are two points of $\tilde{\mathcal{M}}$, then there exists a Mather measure m_0 whose support contains (x,v) and a measure m_0' whose support contains (x',v'). But then the measure $(m_0+m_0')/2$ is clearly a Mather measure whose support contains $\{(x,v),(x',v')\}$ and is itself included in the graph of a K-Lipschitz vectorfield. Assuming that x and x' lie in the image $\theta(B_1)$ of a common chart, see appendix, so that $(x,v)=d\theta(X,V)$ and $(x',v')=d\theta(X',V')$, we obtain

$$||V - V'|| \le K||x - x'||.$$

It follows that the restriction to $\tilde{\mathcal{M}}$ of the canonical projection $TM \longrightarrow M$ is a bi-Lipschitz homeomorphism, or equivalently that the set $\tilde{\mathcal{M}}$ is contained in the graph of a Lipschitz vectorfield. \square

A Notations and standing conventions

- M is a compact manifold of dimension d, and $\pi:TM\longrightarrow M$ is the canonical projection.
- We denote by $\tau: TM \times [0,T] \longrightarrow [0,T]$ or $M \times [0,T] \longrightarrow [0,T]$ the projection on the second factor.
- If N is any separable, complete, locally compact metric space (for example M, $M \times [0, T]$, TM or $TM \times [0, T]$)) the sets $\mathcal{B}_1(N) \subset \mathcal{B}_+(N) \subset \mathcal{B}(N)$ are respectively the set of Borel probability measures, non-negative Borel finite measures, and finite Borel signed measures. If $C_c(N)$ is the set of continuous compactly supported functions on N, endowed with the topology of uniform convergence, then the space $\mathcal{B}(N)$ is identified with the set of

continuous linear forms on $C_c(N)$ by the Riesz theorem. We will always endow the space $\mathcal{B}(N)$ with the weak-* topology that we will also call the weak topology. Note that the set $\mathcal{B}_1(N)$ is compact if N is. Prohorov's theorem states that a sequence of probability measures $P_n \in \mathcal{B}_1(N)$ has a subsequence converging in $\mathcal{B}_1(N)$ for the weak-* topology if for all $\epsilon > 0$ there exists a compact set K_{ϵ} such that $P_n(N - K_{\epsilon}) \leq \epsilon$ for all $n \in \mathbb{N}$. See e.g. [39, 17, 10].

• Given two manifolds N and N', a Borel application $F: N \longrightarrow N'$, and a measure $\mu \in \mathcal{B}(N)$, we define the push-forward $F_{\sharp}\mu$ of μ by F as the unique measure on N' which satisfies

$$F_{\sharp}\mu(B) = \mu(F^{-1}(B))$$

for all Borel set $B \in N$, or equivalently

$$\int_{N'} f d(F_{\sharp}\mu) = \int_{N} f \circ F d\mu$$

for all continuous function $f: N' \longrightarrow \mathbb{R}$.

• A family $\mu_t, t \in [0, T]$ of measures in $\mathcal{B}(N)$ is called measurable if the map $t \longmapsto \int_N f_t d\mu_t$ is Borel measurable for each $f \in C_c(N \times [0, T])$. We define the measure $\mu_t \otimes dt$ on $N \times [0, T]$ by

$$\int_{N\times[0,T]} f d(\mu_t \otimes dt) = \int_0^T \int_N f_t d\mu_t dt$$

for each $f \in C_c(N \times [0,T])$. The well-known desintegration theorem states that, if μ is a measure on $N \times [0,T]$ such that the projected measure on [0,T] is the Lebesgue measure dt, then there exists a measurable family of measures μ_t on N such that $\mu = \mu_t \otimes dt$.

- The set $\mathcal{K}(\mu_0, \mu_T)$ of transport plans is defined in section 1.2.
- The set $\mathcal{I}(\mu_0, \mu_T)$ of initial transport measures is defined in section 2.1.
- The set $\mathcal{M}(\mu_0, \mu_T)$ of transport measures is defined in section 2.1.
- The set $\mathcal{C}(\mu_0, \mu_T)$ of transport currents is defined in section 2.2.
- We fix, once and for all, a finite atlas Θ of M, formed by charts $\theta: B_5 \longrightarrow M$, where B_r is the open ball of radius r centered at zero in \mathbb{R}^d . We assume in addition that the sets $\theta(B_1), \theta \in \Theta$ cover M.
- We say that a vectorfield $X: M \longrightarrow TM$ is K-Lipschitz if, for each chart $\theta \in \Theta$, the mapping $\Pi \circ (d\theta)^{-1} \circ X \circ \theta : B_5 \longrightarrow \mathbb{R}^d$ is K-Lipschitz on B_1 , where Π is the projection $B_5 \times \mathbb{R}^d \longrightarrow \mathbb{R}^d$.
- We mention the following results which are used through the paper: There exists a constant C such that, if A is a subset of M, and $X_A:A\longrightarrow TM$ is a K-Lipschitz vectorfield, then there exists a CK-Lipschitz vectorfield X on M which extends X_A . In addition, if A is a subset of $M\times [0,T]$ and $X_A:A\longrightarrow TM$ is a K-Lipschitz vectorfield, then there exists a CK-Lipschitz vectorfield X on $M\times [0,T]$ which extends X_A . If A is a compact subset of $M\times [0,T]$ and $X_A:A\cap M\times [0,T]$ which extends X_A . If X is a locally Lipschitz vectorfield (which is X is X is a locally Lipschitz on X is a locally Lipschitz vectorfield (which is X is a locally Lipschitz on X is a locally Lipschitz on X is a locally Lipschitz (which is X is X is a locally Lipschitz on X is a local lipschitz on X is

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