

Limit Theorems for Optimal Mass Transportation

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Abstract

The optimal mass transportation was introduced by Monge some 200 years ago and is, today, the source of large number of results in analysis, geometry and convexity. Here I investigate a new, surprising link between optimal transformations obtained by different Lagrangian actions on Riemannian manifolds. As a special case, for any pair of non-negative measures λ^+, λ^- of equal mass

$$W_1(\lambda^-, \lambda^+) = \lim_{\varepsilon \rightarrow 0} \varepsilon^{-1} \inf_{\mu} W_p(\mu + \varepsilon\lambda^-, \mu + \varepsilon\lambda^+)$$

where $W_p, p \geq 1$ is the Wasserstein distance and the infimum is over the set of probability measures in the ambient space.

1 Introduction

The Wasserstein metric W_p ($\infty > p \geq 1$) is a useful distance on the set of positive Borel measures on metric spaces. Given a metric space (M, D) and a pair of positive Borel measures λ^\pm on M satisfying $\int_M d\lambda^+ = \int_M d\lambda^-$:

$$W_p(\lambda^+, \lambda^-) := \inf_{\pi} \left\{ \left[\int_M \int_M D^p(x, y) d\pi(x, y) \right]^{1/p} ; \pi \in \mathcal{P}(\lambda^+, \lambda^-) \right\}, \quad (1.1)$$

where $\mathcal{P}(\lambda^+, \lambda^-)$ stands for the set of all positive Borel measures on $M \times M$ whose M -marginals are λ^+, λ^- .

Under fairly general conditions (e.g if M is compact), a minimizer $\pi^0 \in \mathcal{P}(\lambda^+, \lambda^-)$ of (1.1) exists. Such minimizers are called *optimal plans*. I'll assume in this paper that M is a compact Riemannian manifold and D is a metric related (but not necessarily identical) to the geodesic distance.

If in addition λ^+ satisfies certain regularity conditions, the optimal measure π^0 is supported on a graph of a Borel mapping $\Psi : M \rightarrow M$. By some abuse of notation we call a Borel map Ψ an *optimal plan* if it is a minimizer of

$$W_p(\lambda^+, \lambda^-) = \inf_{\Phi} \left\{ \left[\int D^p(x, \Phi(x)) d\lambda^+ \right]^{1/p} ; \Phi_{\#}\lambda^+ = \lambda^- \right\}$$

(see Section 1.2-4 for notation).

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The metric W_p , $p \geq 1$ is a metrization of the weak topology $C^*(M)$ on positive Borel measures. In particular, it is continuous in the weak topology. Thus, it is possible to approximate $W_p(\lambda^+, \lambda^-)$ (and the corresponding optimal plan) by $W_p(\lambda_N^+, \lambda_N^-)$ on the set of *atomic measures*

$$\lambda_N^\pm \in \mathcal{M}^{+,N} := \left\{ \mu = \sum_{i=1}^N m_i \delta_{(x_i)} \quad , m_i \geq 0, \quad x_i \in M \right\} \quad , N \rightarrow \infty \quad (1.2)$$

reducing (1.1) into a *finite-dimensional linear programming* on the set of non-negative $N \times N$ matrices $\{\mathcal{P}_{i,j}\}$ subjected to linear constraints.

There is, however, a sharp distinction between the case $p > 1$ and $p = 1$. If $p > 1$ then the optimal plan π^0 is unique (for regular λ^+). This is, in general, *not the case* for $p = 1$. Another distinctive feature of the case $p = 1$ is its "pinning property": The distance W_1 depends only on the difference $\lambda := \lambda^+ - \lambda^-$. This is manifested by the alternative, dual formulation of W_1 :

$$W_1(\lambda) = \sup_{\phi} \left\{ \int \phi d\lambda \quad ; \quad \|\phi\|_{Lip} \leq 1 \right\} \quad (1.3)$$

where $\|\phi\|_{Lip} := \sup_{x \neq y \in M} (\phi(x) - \phi(y)) / D(x, y)$.

The optimal potential ϕ yields some partial information on the optimal plan Ψ (if exists). In particular, $\nabla\phi(x)$, whenever exists, only indicates *the direction* of the optimal plan. For example, if the metric D is Euclidean, then $\Psi(x) = x + t(x)\nabla\phi(x)$ for some unknown $t(x) \in \mathbb{R}^+$. This is in contrast to the case $p > 1$ where a dual variational formulation, analogous to (1.3), yields the *complete information* on the optimal plan Ψ in terms of the gradient of some potential ϕ .

In this paper I consider an object called the p -Wasserstein distance ($p > 1$) of λ^+ to λ^- , *conditioned on a probability measure μ* :

$$W^{(p)}(\lambda||\mu) := \sup_{\phi} \left\{ \int \phi d\lambda \quad ; \quad \int |\nabla\phi|^q d\mu \leq 1 \right\} \quad (1.4)$$

where $q = p/(p-1)$.

The first result is

$$W_1(\lambda) = \min_{\mu} \left\{ W^{(p)}(\lambda||\mu) \quad ; \quad \int d\mu = 1 \right\} \quad , (p > 1) \quad (1.5)$$

The problem associated with (1.5) is related to *shape optimization*, see [7]. In addition, the minimizer μ in (1.5) and the corresponding maximizer ϕ in (1.4) or (1.3) play an important rule in the L_1 theory of transport [12]. In fact, the optimal ϕ is, in general, a Lipschitz function which is differentiable μ a.e. and satisfies $|\nabla\phi| = 1$ μ a.e. The minimal measure μ is called a *transport measure*. It verifies the weak form of the continuity equation which, under the current notation, takes the form

$$\nabla \cdot (\mu \nabla \phi) = \frac{\lambda}{W_1(\lambda)} .$$

The transport measure yields an additional information on the optimal plan Ψ along the *transport rays* which completes the information included in $\nabla\phi$ [12]. In the context of shape optimization it is related to the optimal distribution of conducting material [7]. See also [19], [23], [24].

The evaluation of the transport measure μ is therefore an important object of study. It is tempting to approximate the transport measure as a minimizer of (1.5) on a restricted finite space, e.g. for $\mu \in \mathcal{M}^{+,N}$ as defined in (1.2).

However, this cannot be done. Unlike W_p , $W^{(p)}(\lambda||\mu)$ is *not* continuous in the weak topology of C^* on Borel measures with respect to both μ and λ . Indeed, it follows easily that $W^{(p)}(\lambda||\mu) = \infty$ for any atomic measure μ .

The second result of this paper is

$$W^{(p)}(\lambda||\mu) = \lim_{n \rightarrow \infty} nW_p(\mu + \lambda^+/n, \mu + \lambda^-/n) \quad (1.6)$$

Here the limit is in the sense of Γ convergence. A somewhat stronger result is obtained if we take the infimum over all probability measures μ :

$$W_1(\lambda) = \lim_{n \rightarrow \infty} n \min_{\mu} W_p(\mu + \lambda^+/n, \mu + \lambda^-/n) \quad (1.7)$$

where the convergence is, this time, pointwise in λ .

The importance of (1.6, 1.7) is that $W^{(p)}(\lambda||\mu)$ can now be approximated by a *weakly continuous function*

$$W_n^{(p)}(\lambda^+, \lambda^-||\mu) := nW_p(\mu + \lambda^+/n, \mu + \lambda^-/n) .$$

Suppose μ_0 is a unique minimizer of (1.5). If μ_n is a minimizer of $W_n^{(p)}(\lambda^+, \lambda^-||\mu)$ then the sequence $\{\mu_n\}$ must converge to the transport measure μ_0 . In contrast to $W^{(p)}$, $W_n^{(p)}$ is *continuous* in the C^* topology with respect to μ . Hence μ_n can be approximated by atomic measures $\mu_n^N \in \mathcal{M}^{+,N}$ (1.2). In particular a transport measure can be approximated by a finite *points allocation* obtained by minimizing $W_n^{(p)}$ on $\mathcal{M}^{+,N}$ for a sufficiently large n and N .

The results (1.5- 1.7) can be extended to the case where the cost D^p on $M \times M$ is generalized into an action function on a Riemannian manifold $M \times M$, induced by a Lagrangian function $l : TM \rightarrow \mathbb{R}$. This point of view reveals some relations with the *Weak KAM Theory* dealing with invariant measures of Lagrangian flows on manifolds.

1.1 Overview

Section 2 review the necessary background for the Weak KAM and its relation to optimal transport. Section 3 state the main results (Theorems 1-4), which correspond to (1.5- 1.7) for homogeneous Lagrangian on $M \times M$. Section 4 presents the proof of the first of the main results which generalizes (1.4). Finally, Section 5 contains the proofs of the other main results which generalize (1.6, 1.7).

1.2 Standing notations and assumptions

1. (M, g) is a compact, Riemannian Manifold and $D : M \times M \rightarrow \mathbb{R}^+$ is the geodesic distance.
2. TM (res. T^*M) the tangent (res. cotangent) bundle of M . The duality between $v \in T_x M$ and $p \in T_x^* M$ is denoted by $\langle \xi, v \rangle \in \mathbb{R}$. The projection $\Pi : TM \rightarrow M$ is the trivialization $\Pi(x, v) = x$. Likewise $\Pi^* : T^*M \rightarrow M$ is the trivialization $\Pi^*(x, \xi) = x$.
3. For any topological space X , $\mathcal{M}(X)$ is the set of Borel measures on X , $\mathcal{M}_0(X) \subset \mathcal{M}(X)$ the set of such measures which are perpendicular to the constants. $\mathcal{M}^+(X) \subset \mathcal{M}(X)$ the set of all non-negative measures in \mathcal{M} , and $\mathcal{M}_1^+(X) \subset \mathcal{M}^+(X)$ the set of normalized (probability) measures. If $X = M$, the parameter X is usually omitted.
4. A Borel map $\Phi : X_1 \rightarrow X_2$ induces a mapping $\Phi_{\#} : \mathcal{M}^+(X_1) \rightarrow \mathcal{M}^+(X_2)$ via

$$\Phi_{\#}(\mu_1)(A) = \mu_1(\Phi^{-1}(A))$$

for any Borel set $A \subset X_2$.

5. For any $x, y \in M$ let $\mathcal{K}_{x,y}^T$ be the set of all absolutely continuous paths $z : [0, T] \rightarrow M$ connecting x to y , that is, $z(0) = x$, $z(T) = y$.
6. Given $\mu_1, \mu_2 \in \mathcal{M}^+$, the set $\mathcal{P}(\mu_1, \mu_2)$ is defined as all the measures $\Lambda \in \mathcal{M}^+(M \times M)$ such that $\pi_{1,\#}\Lambda = \mu_1$ and $\pi_{2,\#}\Lambda = \mu_2$, where $\pi_i : \mathcal{M} \times \mathcal{M} \rightarrow \mathcal{M}$ defined by $\pi_1(x, y) = x$, $\pi_2(x, y) = y$.
7. An hamiltonian function $h \in C^2(T^*M; \mathbb{R})$ is assumed to be strictly convex and super-linear in ξ on the fibers T_x^*M , uniformly in $x \in M$, that is

$$h(x, \xi) \geq -C + \hat{h}(\xi) \quad \text{where} \quad \lim_{\|\xi\| \rightarrow \infty} \hat{h}(\xi)/\|\xi\| = \infty \quad .$$

The Lagrangian $l : TM \rightarrow \mathbb{R}$ is obtained by Legendre duality

$$l(x, v) = \sup_{\xi \in T_x^* M} \langle \xi, v \rangle - h(x, \xi)$$

satisfies $l \in C^2(TM; \mathbb{R})$, and is super linear on the fibers of $T_x M$ uniformly in x .

8. $Exp_{(l)} : TM \times \mathbb{R} \rightarrow M$ is the flow due to the Lagrangian l on M , corresponding to the Euler-Lagrange equation

$$\frac{d}{dt} l_v = l_x \quad .$$

For each $t \in \mathbb{R}$, $Exp_{(l)}^{(t)} : TM \rightarrow M$ is the exponential map at time t .

2 Background

The weak version of Mather's theory [20] deals with minimal invariant measures of Lagrangians, and the corresponding Hamiltonians defined on a manifold M . In this theory the concept of an orbit $\mathbf{z} = \mathbf{z}(t) : \mathbb{R} \rightarrow M$ is replaced by that of a *closed probability measure* on TM :

$$\mathcal{M}_0^c := \left\{ \nu \in \mathcal{M}_1^+(TM) ; \int_{TM} l(x, v) d\nu(x, v) < \infty , \int_{TM} \langle d\phi, v \rangle d\nu = 0 \text{ for any } \phi \in C^1(M) \right\} . \quad (2.1)$$

A minimal (or Mather) measure $\nu_M \in \mathcal{M}_0^c$ is a minimizer of

$$\inf_{\nu \in \mathcal{M}_0^c} \int_{TM} l(x, v) d\nu(x, v) := -\underline{E} \quad (2.2)$$

It can be shown ([2], [18], [3]) that any minimizer of (2.2) is invariant under the flow induced by the Euler-Lagrange equation on TM :

$$\frac{d}{dt} \nabla_{\dot{x}} l(x, \dot{x}) = \nabla_x l(x, \dot{x}) . \quad (2.3)$$

There is also a dual formulation of (2.2) [17], [29]:

$$\sup_{\mu \in \mathcal{M}_1^+} \inf_{\phi \in C^1(M)} \int_M h(x, d\phi) d\mu = \underline{E} , \quad (2.4)$$

where the maximizer μ_M is the projection of a Mather measure ν_M on M . The ground energy level \underline{E} , common to (2.2, 2.4), admits several equivalent definitions. Evans and Gomes ([11] [13] [14]) defined \underline{E} as the *effective hamiltonian value*

$$\underline{E} := \inf_{\phi \in C^1(M)} \sup_{x \in M} h(x, d\phi) ,$$

while the PDE approach to the WKAM theory ([16], [17]) defines \underline{E} as the minimal $E \in \mathbb{R}$ for which the Hamilton-Jacobi equation $h(x, d\phi) = E$ admits a viscosity sub-solution on M . Alternatively \underline{E} is the *only* constant for which $h(x, d\phi) = \underline{E}$ admits a viscosity solution [15]. There are other, equivalent definitions of \underline{E} known in the literature. We shall meet some of them below.

Example 2.1. *i) $l = l_K := |v|^p / (p - 1)$ where $p > 1$. Here $\underline{E} = 0$ and μ_M is the volume induced by the metric g .*

ii) $l(x, v) = (1/2)|v|^2 - V(x)$ where $V \in C^2(M)$ (mechanical Lagrangian) . Then $\underline{E} = \max_{x \in M} V(x)$ and μ_M of (2.4) is supported at the points of maxima of V .

iii) $l(x, v) = l_K(v - \mathbf{W}(x))$ where \mathbf{W} is a section in TM . Then (2.2) implies $\underline{E} \leq 0$. In fact, it can be shown that $\underline{E} = 0$ for any choice of \mathbf{W} .

iv) In general, if \mathbf{P} is in the first cohomology of M ($\mathbf{H}^1(M)$) then $l \mapsto l(x, v) - \langle \mathbf{P}, v \rangle$ induced the hamiltonian $h \mapsto h(x, \xi + \mathbf{P})$ and $\underline{E} = \alpha(\mathbf{P})$ corresponds to the celebrated Mather (α) function [20] on the cohomology $\mathbf{H}^1(M)$. See also [27].

The Monge problem of mass transportation, on the other hand, has a much longer history. Some years before the the French revolution, Monge (1781) proposed to consider the minimal cost of transporting a given mass distribution to another, where the cost of transporting a unit of mass from point x to y is prescribed by a function $C(x, y)$. In modern language, the Monge problem on a manifold M is described as follows: Given a pair of Borel probability measures μ_0, μ_1 on M , consider the set $\mathcal{K}(\mu_0, \mu_1)$ of all Borel mappings $\Phi : M \rightarrow M$ transporting μ_0 to μ_1 , i.e

$$\Phi \in \mathcal{K}(\mu_0, \mu_1) \iff \Phi_{\#}\mu_0 = \mu_1$$

and look for the one which minimize the *transportation cost*

$$\mathcal{C}(\mu_0, \mu_1) := \inf_{\Phi} \left\{ \int_M C(x, \Phi(x)) d\mu_0(x) \ ; \ \Phi \in \mathcal{K}(\mu_0, \mu_1) \right\} . \quad (2.5)$$

In this generality, the set $\mathcal{K}(\mu_0, \mu_1)$ can be empty if, e.g., μ_0 contains an atomic measure while μ_1 does not, so $\mathcal{C}(\mu_0, \mu_1) = \infty$ in that case. In 1942, Kantorovich proposed a relaxation of this deterministic definition of the Monge cost. Instead of the (very nonlinear) set $\mathcal{K}(\mu_0, \mu_1)$, he suggested to consider the set $\mathcal{P}(\mu_0, \mu_1)$ defined in section 1.2-(6). Then, the definition of the Monge metric is relaxed into the linear optimization

$$\mathcal{C}(\mu_0, \mu_1) = \min_{\Lambda} \left\{ \int_{M \times M} C(x, y) d\Lambda(x, y) \ ; \ \Lambda \in \mathcal{P}(\mu_0, \mu_1) \right\} . \quad (2.6)$$

Example 2.2. The Wasserstein distance W_p ($p \geq 1$) is obtained by the power p of the metric D induced by the Riemannian structure:

$$W_p(\mu_0, \mu_1) = \min_{\Lambda} \left\{ \left[\int_{M \times M} D^p(x, y) d\Lambda(x, y) \right]^{1/p} \ ; \ \Lambda \in \mathcal{P}(\mu_0, \mu_1) \right\} \quad (2.7)$$

The advantage of this relaxed definition is that $\mathcal{C}(\mu_0, \mu_1)$ is always finite, and that a minimizer of (2.6) always exists by the compactness of the set $\mathcal{P}(\mu_0, \mu_1)$ in the weak topology $C^*(M \times M)$. If μ_0 contains no atomic points then it can be shown that $\mathcal{C}(\mu_0, \mu_1)$'s given by (2.5) and (2.6) coincide [1].

The theory of Monge-Kantorovich (M-K) was developed in the last few decades in a countless number of publications. For updated reference see [12], [28].²

Returning now to WKAM, it was observed by Bernard and Buffoni ([4][5]- see also [29]) that the minimal measure and the ground energy can be expressed in terms of the M-K problem subjected to the cost function induced by the Lagrangian (recall section 1.2-5)

$$C_T(x, y) := \inf_{\mathbf{z}} \left\{ \int_0^T l(\mathbf{z}(s); \dot{\mathbf{z}}(s)) ds \ , \ \mathbf{z} \in \mathcal{K}_{x,y}^T \right\} , T > 0 . \quad (2.8)$$

²By convention, the name "Monge problem" is reserved for the metric cost, while "Monge-Kantorovich problem" is usually referred to general cost functions

Then

$$\mathcal{C}_T(\mu) := \mathcal{C}_T(\mu, \mu) = \min_{\Lambda} \left\{ \int_{M \times M} C_T(x, y) d\Lambda(x, y) \ ; \ \Lambda \in \mathcal{P}(\mu, \mu) \right\}$$

and

$$\min_{\mu} \{ \mathcal{C}_T(\mu) \ ; \ \mu \in \mathcal{M}_1^+ \} = -T\underline{E} \tag{2.9}$$

where the minimizers of (2.9) coincide, for any $T > 0$, with the projected Mather measure μ_M maximizing (2.4) [5]. The action C_T induces a metric on the manifold M :

$$(x, y) \in M \times M \mapsto D_E(x, y) = \inf_{T > 0} C_T(x, y) + TE . \tag{2.10}$$

Example 2.3.

i) For $l(x, v) = |v|^p/(p-1)$, $p > 1$ we get $C_T(x, y) = D(x, y)^p/(p-1)T^{p-1}$ while $D_E(x, y) = pE^{1-1/p}D_g(x, y)/(p-1)$ if $E \geq 0$, $D_E(x, y) = -\infty$ if $E < 0$.

ii) $l(x, v) = (1/2)|v|^2 - V(x)$ where $V \in C^2(M)$ (mechanical Lagrangian) . Then $D_E(x, y)$ is the geodesic distance induced by conformal equivalent metric $(M, (E - V)g)$ on M , where $E \geq \underline{E} = \sup_M V$.

It is not difficult to see that either $D_E(x, x) = 0$ for any $x \in M$, or $D_E(x, y) = -\infty$ for any $x, y \in M$. In fact, it follows ([22], [10]) that $D_E(x, y) = -\infty$ for $E < \underline{E}$ and $D_E(x, x) = 0$ for $E \geq \underline{E}$ and any $x, y \in M$.

Let now $\lambda^+, \lambda^- \in \mathcal{M}^+$ where $\lambda := \lambda^+ - \lambda^- \in \mathcal{M}_0$, that is $\int_M d\lambda = 0$. Let

$$\mathcal{D}_E(\lambda) := \mathcal{D}_E(\lambda^+, \lambda^-) = \min_{\Lambda} \left\{ \int_{M \times M} D_E(x, y) d\Lambda(x, y) \ ; \ \Lambda \in \mathcal{P}(\lambda) \right\} \quad (2.11)$$

be the Monge distance of λ^+ and λ^- with respect to the metric D_E . There is a dual formulation of \mathcal{D}_E as follows: Consider the set \mathcal{L}_E of D_E Lipschitz functions on M :

$$\mathcal{L}_E := \{ \phi \in C(M) \ ; \ \phi(x) - \phi(y) \leq D_E(x, y) \ \forall x, y \in M \} \quad (2.12)$$

Then (see, e.g [12], [26])

$$\mathcal{D}_E(\lambda) = \max_{\phi} \left\{ \int_M \phi d\lambda \ ; \ \phi \in \mathcal{L}_E \right\} \quad (2.13)$$

3 Description of the main results

The object of this paper is to establish some relations between the action C_T and a modified action \widehat{C}_T defined below.

3.1 Unconditional action

For given $\lambda \in \mathcal{M}_0$ we generalize (2.1) into

$$\mathcal{M}_\lambda := \left\{ \nu \in \mathcal{M}_1^+(TM) \ ; \ \int_{TM} l(x, v) d\nu(x, v) < \infty \ ; \ \int_{TM} \langle d\phi, v \rangle d\nu = \int_M \phi d\lambda \ \text{for any } \phi \in C^1(M) \right\} \quad (3.1)$$

and define

$$\widehat{C}(\lambda) := \inf_{\nu} \left\{ \int_{TM} l(x, v) d\nu(x, v) \ ; \ \nu \in \mathcal{M}_\lambda \right\} . \quad (3.2)$$

The modified action $\widehat{C}_T : \mathcal{M}_0 \rightarrow \mathbb{R} \cup \{\infty\}$, $T > 0$ have several equivalent definitions as given in Theorem 1 below:

Theorem 1. *The following definitions are equivalent:*

1. $\widehat{C}_T(\lambda) := T\widehat{C}(\frac{\lambda}{T})$.

$$2. \widehat{\mathcal{C}}_T(\lambda) := \min_{\mu} \sup_{\phi} \left\{ \int_M -Th(x, d\phi) d\mu + \phi d\lambda \ ; \ \mu \in \mathcal{M}_1^+, \phi \in C^1(M) \right\} .$$

$$3. \widehat{\mathcal{C}}_T(\lambda) := \max_{E \geq \underline{E}} [\mathcal{D}_E(\lambda) - ET] .$$

In addition if $T_c := D_{\underline{E}}^+(\lambda) < \infty$ then for $T \geq T_c$,

$$\widehat{\mathcal{C}}_T(\lambda) = \widehat{\mathcal{C}}_{T_c}(\lambda) - T\underline{E} .$$

In that case the minimizer $\mu_{\lambda}^T \in \mathcal{M}_1^+$ of (3), $T > T_c$ is given by

$$\mu_{\lambda}^T = \frac{T_c}{T} \mu_{\lambda}^{T_c} + \left(1 - \frac{T_c}{T}\right) \mu_M ,$$

where μ_M is the projected Mather measure.

Remark 3.1. Note that $\mathcal{D}_E(\lambda)$ (2.11, 2.13) is a monotone non-decreasing and concave function of E while $\mathcal{D}_{\underline{E}}(\lambda) > -\infty$ by definition. Hence the right-derivative of $\mathcal{D}_{\underline{E}}^+(\lambda)$ as a function of E is defined and positive (possibly $+\infty$ at $E = \underline{E}$).

Remark 3.2. A special case of Theorem 1 was introduced in [30].

For the next result we need a two technical assumptions:

H₁ There exists a sequence of smooth, positive mollifiers $\delta_{\varepsilon} : M \times M \rightarrow \mathbb{R}^+$ such that, for any $\phi \in C^0(M)$ (res. $\phi \in C^1(M)$)

$$\lim_{\varepsilon \rightarrow 0} \delta_{\varepsilon} * \phi = \phi$$

where the convergence is in $C^0(M)$ (res. $C^1(M)$) and for any $\varepsilon > 0$ and $\phi \in C^1(M)$

$$\delta_{\varepsilon} * d\phi = d(\delta_{\varepsilon} * \phi) .$$

H₂ For any $(x, p) \in T^*M$ and $\varepsilon > 0$ there exists $\delta > 0$ such that $h(x, \xi) - h(y, \xi_y) \leq \varepsilon(h(x, \xi) + 1)$ provided $D(x, y) < \delta$. Here ξ_y is obtained by parallel translation of (x, ξ) to y .

Remark 3.3. **H₁** holds for homogeneous spaces, e.g the flat d -torus $\mathbb{R}^d/\mathbb{Z}^n$ or the sphere $\mathbb{S}^{d-1} = SO(d)/SO(1)$.

H₂ holds, in particular, for any mechanical hamiltonian with continuous potential.

Theorem 2. Assume **H₁** + **H₂**. For any $\lambda = \lambda^+ - \lambda^-$ where $\lambda^{\pm} \in \mathcal{M}_1^+$,

$$\widehat{\mathcal{C}}_T(\lambda) = \lim_{\varepsilon \rightarrow 0} \min_{\mu \in \mathcal{M}_1^+} \varepsilon^{-1} \mathcal{C}_{\varepsilon T}(\mu + \varepsilon\lambda^-, \mu + \varepsilon\lambda^+) .$$

As an application of Theorem 2 we may consider the case where the lagrangian l is homogeneous with respect to a Riemannian metric $g(x)$:

Example 3.1. If $l(x, v) = |v|^p/(p-1)$ where $p > 1$. Then $C_T(x, y) = \frac{D^p(x, y)}{(p-1)T^{p-1}}$ while $D_E(x, y) = \frac{p}{p-1}E^{(p-1)/p}D(x, y)$ and $\underline{E} = 0$. It follows that

$$\widehat{C}_T(\lambda) = \frac{W_1^p(\lambda)}{(p-1)T^{p-1}}, \quad \varepsilon^{-1}C_{\varepsilon T}(\mu + \varepsilon\lambda^+, \mu + \varepsilon\lambda^-) = \frac{W_p^p(\mu + \varepsilon\lambda^+, \mu + \varepsilon\lambda^-)}{(p-1)T^{p-1}\varepsilon^p} \quad (3.3)$$

where the Wasserstein distance W_p is defined in (2.7). Hence, by Theorem 1 and Theorem 2

$$W_1(\lambda) = \lim_{\varepsilon \rightarrow 0} \varepsilon^{-1} \inf_{\mu \in \mathcal{M}_1^+} W_p(\mu + \varepsilon\lambda^-, \mu + \varepsilon\lambda^+).$$

Remark 3.4. The optimal transport description of the weak KAM theory (2.9) can be considered as a special case of Theorem 2 where $\lambda = 0$. Indeed $\inf_{\mu \in \mathcal{M}_1^+} \varepsilon^{-1}C_{\varepsilon T}(\mu, \mu) = -T\underline{E}$ by (2.9). On the other hand, since $\mathcal{D}_E(0) = 0$ for any $E \geq \underline{E}$ it follows that $T_c = 0$, hence $\widehat{C}_{T_c}(0) = 0$ so $\widehat{C}_T(0) = -T\underline{E}$ as well by the last part of Theorem 1.

3.2 Conditional action

There is also an interest in the definition of action (and metric distance) conditioned with a given probability measure $\mu \in \mathcal{M}_1^+$. We introduce these definitions and reformulate parts of the main results Theorems 1-2 in terms of these.

For a given $\mu \in \mathcal{M}_1^+$ and $E \geq \underline{E}$, let

$$\mathcal{H}_E(\mu) := \left\{ \phi \in C^1(M) ; \int_M h(x, d\phi) d\mu \leq E \right\}. \quad (3.4)$$

In analogy with (2.13) we define the μ -conditional metric on $\lambda \in \mathcal{M}_0$:

$$\mathcal{D}_E(\lambda \| \mu) := \sup_{\phi} \left\{ \int_M \phi d\lambda ; \phi \in \mathcal{H}_E(\mu) \right\}. \quad (3.5)$$

The conditioned, modified action with respect to $\mu \in \mathcal{M}_1^+$ is defined in analogy with Theorem 1 (2, 3)

$$\widehat{C}_T(\lambda \| \mu) := \max_{E \geq \underline{E}} \mathcal{D}_E(\lambda \| \mu) - ET \equiv \sup_{\phi \in C^1(M)} \int_M -Th(x, d\phi) d\mu + \phi d\lambda. \quad (3.6)$$

Example 3.2. As in Example 3.1, $l(x, v) = |v|^p/(p-1)$ implies $h(\xi) = q^{-q}|\xi|^q$ where $q = p/(p-1)$. Then (3.4, 3.5) is related to (1.4), that is $W_1^{(p)}(\lambda \| \mu) = \mathcal{D}_E(\lambda \| \mu)$ where $E = q^{-q}$ or

$$\mathcal{D}_E(\lambda \| \mu) = qE^{1/q}W_1^{(p)}(\lambda \| \mu), \quad \widehat{C}_T(\lambda \| \mu) = \frac{q-1}{T^{1/(q-1)}} \left(W_1^{(p)}(\lambda \| \mu) \right)^p \quad (3.7)$$

Remark 3.5. It seems there is a relation between this definition and the tangential gradient [6]. There are also possible applications to optimal network and irrigation theory, where one wishes to minimize $D(\lambda \| \mu)$ over some constrained set of $\mu \in \mathcal{M}_1^+$ (the irrigation network) for a prescribed λ (representing the set of sources and targets). See, e.g. [8], [9] and the ref. within.

The next result is

Theorem 3. For any $\lambda \in \mathcal{M}_0$,

$$\mathcal{D}_E(\lambda) = \min_{\mu \in \mathcal{M}_1^+} \mathcal{D}_E(\lambda \|\mu) , \quad \widehat{\mathcal{C}}_T(\lambda) = \min_{\mu \in \mathcal{M}_1^+} \widehat{\mathcal{C}}_T(\lambda \|\mu) .$$

The analog of Theorem 2 holds for the conditional action as well. However, we can only prove the Γ -convergence in that case. Recall that a sequence of functionals $F_n : \mathbf{X}_n \rightarrow \mathbb{R} \cup \{\infty\}$ is said to Γ -converge to $F : \mathbf{X} \rightarrow \mathbb{R} \cup \{\infty\}$ ($\Gamma - \lim_{n \rightarrow \infty} F_n = F$) if and only if

- (i) $\mathbf{X}_n \subset \mathbf{X}$ for any n .
- (ii) For any sequence $x_n \in \mathbf{X}_n$ converging to $x \in \mathbf{X}$ in the topology of \mathbf{X} ,

$$\liminf_{n \rightarrow \infty} F_n(x_n) \geq F(x) .$$

- (iii) For any $x \in \mathbf{X}$ there exists a sequence $\hat{x}_n \in \mathbf{X}_n$ converging to $x \in \mathbf{X}$ in the topology of \mathbf{X} for which

$$\lim_{n \rightarrow \infty} F_n(\hat{x}_n) = F(x) .$$

In Theorem 4 below the Γ -convergence is related to the special case where $\mathbf{X}_n = \mathbf{X}$:

Theorem 4. Let $\mathbf{X}_n = \mathcal{M}_0 \times \mathcal{M}_1^+ = \mathbf{X}$ and $F_n(\lambda, \mu) := n\mathcal{C}_{T/n}(\mu + \lambda^-/n, \mu + \lambda^+/n)$. Then

$$\widehat{\mathcal{C}}_T(\cdot \|\cdot) = \Gamma - \lim_{n \rightarrow \infty} F_n .$$

From Theorem 4 and Theorem 2 it follows immediately

Corollary 3.1. In addition, if μ_n is a minimizer of F_n in \mathcal{M}_1^+ then any converging subsequence of μ_n , $n \rightarrow \infty$, converges to a minimizer of $\widehat{\mathcal{C}}(\lambda \|\cdot)$ in \mathcal{M}_1^+ .

Finally, we note that (1.7) is a special case of Theorem 4. Using Examples 3.1, 3.2 with $\varepsilon = 1/n$, recalling $(q-1)^{-1} = p-1$ we obtain

Corollary 3.2.

$$W_1(\lambda) = \lim_{n \rightarrow \infty} n \min_{\mu \in \mathcal{M}_1^+} W_p(\mu + \lambda^+/n, \mu + \lambda^-/n)$$

4 Proof of Theorems 1&3

We first show that $\widehat{\mathcal{C}}(\lambda) < \infty$ (recall (3.2)).

Lemma 4.1. For any $\lambda \in \mathcal{M}_0$, $\mathcal{M}_\lambda \neq \emptyset$. In particular, since the Lagrangian l is bounded from below, $\widehat{\mathcal{C}}(\lambda) < \infty$.

Proof. It is enough to show that there exists a compact set $K \subset TM$ and a sequence $\{\lambda_n\} \subset \mathcal{M}_0$ converging weakly to λ such that for each n there exists $\nu_n \in \mathcal{M}_{\lambda_n}$ whose support is contained in K . Indeed, such a set is compact and there exists a weak limit $\nu = \lim_{n \rightarrow \infty} \nu_n$ which satisfies $\lim_{n \rightarrow \infty} \int_M \langle d\phi, v \rangle d\nu_n = \int_M \langle d\phi, v \rangle d\nu$ as well. Hence, if $\phi \in C^1(M)$ then

$$\lim_{n \rightarrow \infty} \int_M \langle d\phi, v \rangle d\nu_n = \int_M \langle d\phi, v \rangle d\nu \quad , \quad \lim_{n \rightarrow \infty} \int_M \phi d\lambda_n = \int_M \phi d\lambda_n .$$

Since $\nu_n \in \mathcal{M}_{\lambda_n}$ we get

$$\int_M \langle d\phi, v \rangle d\nu_n = \int_M \phi d\lambda_n$$

for any n , so the same equality holds for ν as well.

Now, we consider

$$\lambda_n = \alpha_n \sum_{j=1}^n (\delta_{x_j} - \delta_{y_j}) \tag{4.1}$$

where $x_j, y_j \in M$ and $\alpha_n > 0$. For any pair (x_j, y_j) consider a geodesic arc corresponding to the Riemannian metric which connect x to y , parameterized by the arc length: $\mathbf{z}_j : [0, 1] \rightarrow M$ and $|\dot{\mathbf{z}}| = D(x_j, y_j)$ (recall section 1.2-(1)). Then

$$\nu_n := \alpha_n \sum_{j=1}^n \int_0^1 \delta_{x - \mathbf{z}_j(t), v - \dot{\mathbf{z}}_j(t)} dt$$

satisfies for any $\phi \in C^1(M)$

$$\begin{aligned} \int_M \langle d\phi, v \rangle d\nu_n &= \alpha_n \sum_{j=1}^n \int_0^1 \langle d\phi(\mathbf{z}_j(s)), \dot{\mathbf{z}}_j(s) \rangle \dot{\mathbf{z}}_j(t) dt = \alpha_n \sum_{j=1}^n \int_0^1 \frac{d}{dt} \phi(\mathbf{z}_j(s)) dt \\ &= \alpha_n \sum_{j=1}^n [\phi(y_j) - \phi(x_j)] = \int_M \phi d\lambda_n \end{aligned} \tag{4.2}$$

hence $\nu_n \in \mathcal{M}_{\lambda_n}$. Finally, we can certainly find such a sequence λ_n of the form (4.1) which converges weakly to λ . \square

4.1 Point distances and Hamiltonians

For $E \in \mathbb{R}$, let $\sigma_E : TM \rightarrow \mathbb{R}$ the support function of the level surface $h(x, \xi) \leq E$, that is:

$$\sigma_E(x, v) := \sup_{\xi \in T_x^*M} \{ \langle \xi, v \rangle_{(x)} ; h(x, \xi) \leq E \} . \tag{4.3}$$

It follows from our standing assumptions (Section 1.2-7) that σ_E is differentiable as a function of E for any $(x, v) \in TM$. For the following Lemma see e.g. [25].

Recall that

$$D_E(x, y) := \inf_{T > 0} C_T(x, y) + ET \tag{4.4}$$

where C_T as defined in (2.8). Recall also section 1.2-5:

Lemma 4.2. .

$$D_E(x, y) = \inf_{z \in \mathcal{K}_{x,y}^1} \int_0^1 \sigma_E(z(s), \dot{z}(s)) ds . \quad (4.5)$$

Given $x \in M$, let

$$\underline{E} := \inf \{E \in \mathbb{R}; D_E(x, x) > -\infty\} \quad (4.6)$$

For the following Lemma see [21] (also [27]):

Lemma 4.3. \underline{E} is independent of $x \in M$. The definitions (4.6) and (2.2) and (2.4) are equivalent. If $E \geq \underline{E}$ then $D_E(x, y) > -\infty$ for any $x, y \in M$ and, in addition

i) $D_E(x, x) = 0$ for any $x \in M$.

ii) For any $x, y, z \in M$, $D_E(x, z) \leq D_E(x, y) + D_E(y, z)$

From (4.4), Lemma 4.2 and the continuity of σ_E with respect to $E \geq \underline{E}$ we get

Corollary 4.1. If $E \geq \underline{E}$ then for any $x, y \in M$, $D_E(x, y)$ is continuous, monotone non-decreasing and concave as a function of E .

Note that the differentiability of σ_E with respect to E does *not* imply that $D_E(x, y)$ is differentiable for each $x, y \in M$. However, since $D_E(x, y)$ is a concave function of E for each $x, y \in M$, it is differentiable for Lebesgue almost any $E > \underline{E}$. We then obtain by differentiation

Lemma 4.4. If E is a point of differentiability of $D_E(x, y)$ then there exists a geodesic arc $z \in \mathcal{K}_{x,y}^1$ realizing (4.5) such that the E derivative of $D_E(x, y)$ is given by

$$T_E(x, y) := \frac{d}{dE} D_E(x, y) = \int_0^1 \sigma'_E(z(s), \dot{z}(s)) ds , \quad (4.7)$$

where σ'_E is the E derivative of σ_E . Moreover

$$D_E(x, y) = C_{T_E(x,y)}(x, y) + ET_E(x, y) . \quad (4.8)$$

From (4.3) we get $\sigma_E(x, v) \leq |v| \max\{|p| ; h(x, \xi) \leq E\}$. From our standing assumption on h (section 1.2-(7)) and (4.5) we obtain

Lemma 4.5. For any $x, y \in M$ and $E \geq \underline{E}$

$$D_E(x, y) \leq \hat{h}^{-1}(E + C)D(x, y)$$

In particular

$$\lim_{E \rightarrow \infty} E^{-1} D_E(x, y) = 0 \quad (4.9)$$

uniformly on $M \times M$.

Corollary 4.2. For $E \geq \underline{E}$, the set \mathcal{L}_E (2.12) is contained in the set of Lipschitz functions with respect to D , and \mathcal{L}_E is locally compact in $C(M)$.

Given $\phi \in C^1(M)$ let

$$\overline{H}(\phi) := \sup_{x \in M} h(x, d\phi) . \quad (4.10)$$

We extend the definition of \overline{H} to the larger class of Lipschitz functions by the following

Lemma 4.6. *If $\phi \in C^1(M)$ then*

$$\overline{H}(\phi) = \min_{E \geq \phi} \{E; \phi \in \mathcal{L}_E\} ,$$

where \mathcal{L}_E as defined in (2.12).

Proof. First we show that if $\phi \in \mathcal{L}_E \cap C^1(M)$ then $h(x, d\phi) \leq E$ for all $x \in M$. Indeed, for any $x, y \in M$ and any curve $z(\cdot)$ connecting x to y

$$\phi(y) - \phi(x) = \int_0^1 d\phi(z(t)) \cdot \dot{z} dt \leq D_E(x, y) \leq \int_0^1 \sigma_E(z(t), \dot{z}(t)) dt$$

hence $d\phi(x) \cdot v \leq \sigma_E(x, v)$ for any $v \in T_x M$. Then, by definition, $d\phi(x)$ is contained in any supporting half space which contains the set $Q_x(E) := \{\xi \in T_x^* M; h(x, \xi) \leq E\}$. Since this set is convex by assumption, it follows that $d\phi \in Q_x(E)$, so $h(x, d\phi) \leq E$ for any $x \in M$. Hence $\overline{H}(\phi) \leq E$.

Next we show the opposite inequality $h(x, d\phi) \geq E$ for all $x \in M$. Recall (4.8). Then for any $\varepsilon > 0$ we can find $T_\varepsilon > 0$ and $\mathbf{z}_\varepsilon \in \mathcal{K}_{x,y}^{T_\varepsilon}$ so

$$D_E(x, y) \geq \int_0^{T_\varepsilon} l(\mathbf{z}_\varepsilon(t), \dot{\mathbf{z}}_\varepsilon(t)) dt + (E - \varepsilon)T_\varepsilon . \quad (4.11)$$

Next, for a.e $t \in [0, T_\varepsilon]$

$$h(\mathbf{z}_\varepsilon(t), d\phi(\mathbf{z}_\varepsilon(t))) \geq \dot{\mathbf{z}}_\varepsilon(t) \cdot d\phi(\mathbf{z}_\varepsilon(t)) - l(\mathbf{z}_\varepsilon(t), \dot{\mathbf{z}}_\varepsilon(t)) . \quad (4.12)$$

Integrate (4.12) from 0 to T_ε and use $\mathbf{z}_\varepsilon \in \mathcal{K}_{x,y}^{T_\varepsilon}$, (4.11, 4.12) and the definition of \mathcal{L}_E to obtain

$$T_\varepsilon^{-1} \int_0^{T_\varepsilon} h(\mathbf{z}_\varepsilon(t), d\phi(\mathbf{z}_\varepsilon(t))) dt \geq T_\varepsilon^{-1} [\phi(y) - \phi(x)] - T_\varepsilon^{-1} \int_0^{T_\varepsilon} l(\mathbf{z}_\varepsilon(t), \dot{\mathbf{z}}_\varepsilon(t)) dt \geq E - \varepsilon .$$

Hence, the supremum of $h(x, d\phi)$ along the orbit of \mathbf{z}_ε is, at least, $E - \varepsilon$. Since ε is arbitrary, then $\overline{H}(\phi) \geq E$. \square

4.2 Measure distances and Hamiltonians

From Lemma 4.6 and Corollary 4.2 we extend the definition of \overline{H} to the space $Lip(M)$ of Lipschitz functions on M . Let now define for $\lambda \in \mathcal{M}_0$

$$\overline{H}_T^*(\lambda) := \sup_{\phi \in Lip(M)} \left\{ -T\overline{H}(\phi) + \int_M \phi d\lambda \right\} \in \mathbb{R} \cup \{\infty\} . \quad (4.13)$$

Proposition 4.1. *For any $\lambda \in \mathcal{M}_0$*

$$\overline{H}_T^*(\lambda) = \sup_{E \geq \underline{E}} \{ \mathcal{D}_E(\lambda) - TE \} . \quad (4.14)$$

Proof. By definition of \overline{H}^* and Lemma 4.6,

$$\begin{aligned} \overline{H}_T^*(\lambda) &= \sup_{\phi \in \text{Lip}(M)} \left[\int_M \phi d\lambda - T\overline{H}(\phi) \right] = \sup_{\phi \in \text{Lip}(M)} \sup_{E \geq \underline{E}} \left[\int_M \phi d\lambda - TE ; \phi \in \mathcal{L}_E \right] \\ &= \sup_{E \geq \underline{E}} \sup_{\phi \in \text{Lip}(M)} \left[\int_M \phi d\lambda - TE ; \phi \in \mathcal{L}_E \right] = \sup_{E \geq \underline{E}} \{ \mathcal{D}_E(\lambda) - TE \}, \end{aligned} \quad (4.15)$$

where we used the duality relation given by (2.13). \square

Corollary 4.3. \overline{H}_T^* is weakly continuous on \mathcal{M}_0 .

Proof. For each $E \geq \underline{E}$, the Monge-Kantorovich metric $\mathcal{D}_E : \mathcal{M}_0 \rightarrow \mathbb{R}$ is continuous on \mathcal{M}_0 (under weak* topology). Indeed, it is u.s.c. by (2.11) and l.s.c. by the dual formulation (2.13).

Also, for each $\lambda \in \mathcal{M}_1^+$, $\mathcal{D}_E(\lambda)$ is concave and finite in E for $E \geq \underline{E}$. It follows that \mathcal{D} is mutually continuous on $[\underline{E}, \infty[\times \mathcal{M}_0$. From (4.9) we also get that \mathcal{D} is coercive on \mathcal{M}_0 , that is $\lim_{E \rightarrow \infty} E^{-1} \mathcal{D}_E(\lambda) = 0$ locally uniformly on \mathcal{M}_0 . These imply that \overline{H}_T^* is continuous on \mathcal{M}_0 via (4.14). \square

We return now to Corollary 4.1 and Lemma 4.4. It follows that for any countable dense set $A \subset M$ there exists a (possibly empty) set $N \subset]\underline{E}, \infty[$ of zero Lebesgue measure such that $D_E(x, y)$ is differentiable in $E \in]\underline{E}, \infty[- N$, for any $x, y \in A$. Let $\mathcal{M}(A) \subset \mathcal{M}_0$ be the set of all measures in \mathcal{M}_0 which are supported on a finite subset of A , and such that $\lambda(\{x\})$ is rational for any $x \in A$. Again, since $\mathcal{M}(A)$ is countable, it follows by Corollary 4.1 that $\mathcal{D}_E(\lambda)$ is differentiable (as a function of E) for any $\lambda \in \mathcal{M}(A)$ and any $E \in]\underline{E}, \infty[- N$ for a (perhaps larger) set N of zero Lebesgue measure. It is also evident that \mathcal{M}_0 is the weak closure of $\mathcal{M}(A)$.

Lemma 4.7. *For any $\lambda^+ - \lambda^- \equiv \lambda \in \mathcal{M}(A)$ and $E \in]\underline{E}, \infty[- N$, there exists an optimal plan $\Lambda_E^\lambda \in \mathcal{P}(\lambda^+, \lambda^-)$ realizing*

$$\int_{M \times M} D_E(x, y) d\Lambda_E^\lambda(x, y) = \min_{\Lambda \in \mathcal{P}(\lambda^+, \lambda^-)} \int_{M \times M} D_E(x, y) d\Lambda(x, y) \equiv \mathcal{D}_E(\lambda) \quad (4.16)$$

for which

$$\frac{d}{dE} \mathcal{D}_E(\lambda) = \sum_{x, y \in A} \Lambda_E^\lambda(\{x, y\}) T_E(x, y) . \quad (4.17)$$

Proof. Let $E_n \searrow E$. For each n , set $\Lambda_{E_n}^\lambda$ be a minimizer of (4.16) subjected to $E = E_n$. We choose a subsequence so that the limit

$$\Lambda_{E^+}^\lambda(\{x, y\}) := \lim_{n \rightarrow \infty} \Lambda_{E_n}^\lambda(\{x, y\}) \quad (4.18)$$

exists for any $x, y \in A$. Evidently, $\Lambda_{E^+}^\lambda \in \mathcal{P}(\lambda^+, \lambda^-)$ is an optimal plan for (4.16). Next,

$$\mathcal{D}_{E_n}(\lambda) - \mathcal{D}_E(\lambda) \geq \sum_{x, y \in A} \Lambda_{E_n}^\lambda(\{x, y\}) (D_{E_n}(x, y) - D_E(x, y))$$

Divide by $E_n - E > 0$ and let $n \rightarrow \infty$, using (4.18) and (4.7) we get

$$\frac{d}{dE} \mathcal{D}_E(\lambda) \geq \sum_{x, y \in A} \Lambda_{E^+}^\lambda(\{x, y\}) T_E(x, y). \quad (4.19)$$

We repeat the same argument for a sequence $E^n \nearrow E$ for which

$$\Lambda_{E^-}^\lambda(\{x, y\}) := \lim_{n \rightarrow \infty} \Lambda_{E^n}^\lambda(\{x, y\})$$

and get

$$\frac{d}{dE} \mathcal{D}_E(\lambda) \leq \sum_{x, y \in A} \Lambda_{E^-}^\lambda(\{x, y\}) T_E(x, y). \quad (4.20)$$

Again $\Lambda_{E^-}^\lambda$ is an optimal plan as well. If $\Lambda_{E^-}^\lambda = \Lambda_{E^+}^\lambda$ then we are done. Otherwise, define $\Lambda_{E^-}^\lambda$ as a convex combination of $\Lambda_{E^-}^\lambda$ and $\Lambda_{E^+}^\lambda$ for which the equality (4.17) holds due to (4.19, 4.20). \square

Given $x, y \in M$, let E be a point of differentiability of $D_E(x, y)$, and $\mathbf{z}_{x, y}^E : [0, 1] \rightarrow M$ a geodesic arc connecting x, y and realizing (4.7). Then $d\tau_{x, y}^E := \sigma'_E(\mathbf{z}_{x, y}^E, \dot{\mathbf{z}}_{x, y}^E) ds$ is a non-negative measure on $[0, 1]$, and (4.7) reads $T_E(x, y) = \int_0^1 d\tau_{x, y}^E$. Let $\mu_{x, y}^E$ be the measure on M obtained by pushing $\tau_{x, y}^E$ from $[0, 1]$ to M via $\mathbf{z}_{x, y}^E$:

$$\mu_{x, y}^E := (\mathbf{z}_{x, y}^E)_\# \tau_{x, y}^E \in \mathcal{M}^+,$$

that is, for any $\phi \in C(M)$,

$$\int_M \phi d\mu_{x, y}^E := \int_0^1 \phi(\mathbf{z}_{x, y}^E(t)) d\tau_{x, y}^E. \quad (4.21)$$

Definition 4.1. For any $\lambda \in \mathcal{M}(A)$ and $E \in]\underline{E}, \infty[-N$ let

$$\mu_\lambda^E := \sum_{x, y \in A} \Lambda_E^\lambda(\{x, y\}) \mu_{x, y}^E$$

where $\mu_{x, y}^E$ are as given in (4.21) and Λ_E^λ is the particular optimal plan given in Lemma 4.7.

Remark 4.1. Note that $\int_M d\mu_\lambda^E = \mathcal{D}'_E(\lambda)$ for any $\lambda \in \mathcal{M}_0(A)$ and $E \in]\underline{E}, \infty[-N$ by Lemma 4.7, where $\mathcal{D}'_E(\lambda) = (d/dE)\mathcal{D}_E(\lambda)$.

Definition 4.2. For any $\lambda \in \mathcal{M}_0$, $T > 0$, $E(\lambda, T)$ is the maximizer of (4.14), that is

$$\mathcal{D}_{E(\lambda, T)}(\lambda) - TE(\lambda, T) \equiv \overline{H}_T^*(\lambda).$$

By Corollary 4.1 (in particular, the concavity of $\mathcal{D}_E(\lambda)$ with E) we obtain

Lemma 4.8. *If $E(\lambda, T) > \underline{E}$ then*

$$\frac{d^+}{dE} \mathcal{D}_E(\lambda) \Big|_{E=E(\lambda, T)} \leq T \leq \frac{d^-}{dE} \mathcal{D}_E(\lambda, T) \Big|_{E=E(\lambda, T)}$$

where d^+/dE (res. d^-/dE) stands for the right (res. left) derivative. If $E(\lambda, T) = \underline{E}$ then

$$\frac{d^+}{dE} \mathcal{D}_E(\lambda) \Big|_{E=\underline{E}} \leq T .$$

4.3 Proof of Theorem 1 (1 \Leftrightarrow 2)

First we note that it is enough to assume $T = 1$. Consider

$$\mathcal{F}(\mu, \phi) := \int_M -h(x, d\phi) d\mu + \phi d\lambda \quad (4.22)$$

where $\lambda \in \mathcal{M}_0$ is prescribed. Evidently, \mathcal{F} is convex lower semi continuous (l.s.c) in μ on \mathcal{M}_1^+ and concave upper semi continuous (u.s.c) in ϕ on $C^1(M)$. Since \mathcal{M}_1^+ is compact, the Minimax Theorem implies

$$\sup_{\phi \in C^1(M)} \min_{\mu \in \mathcal{M}_1^+} \mathcal{F}(\mu, \phi) = \min_{\mu \in \mathcal{M}_1^+} \sup_{\phi \in C^1(M)} \mathcal{F}(\mu, \phi) . \quad (4.23)$$

Next define

$$\mathcal{G}(\nu, \phi) := \int_{TM} (l(x, v) - \langle d\phi, v \rangle) d\nu + \int_M \phi d\lambda .$$

on $\mathcal{M}_1^+(TM) \times C^1(M)$. Then (recall (3.1))

$$\sup_{\phi \in C^1(M)} \inf_{\nu \in \mathcal{M}_1^+(TM)} \mathcal{G}(\nu, \phi) \leq \inf_{\nu \in \mathcal{M}_\lambda} \int_{TM} l(x, v) d\nu \equiv \widehat{\mathcal{C}}(\lambda) . \quad (4.24)$$

Now

$$\overline{\mathcal{G}}(\nu) := \sup_{\phi \in C^1(M)} \mathcal{G}(\nu, \phi) \equiv \begin{cases} \int_{TM} l(x, v) d\nu & \text{if } \nu \in \mathcal{M}_\lambda \\ \infty & \text{if } \nu \notin \mathcal{M}_\lambda \end{cases} .$$

We recall, again, from the Minmax Theorem that the inequality in (4.24) turns into an equality provided the set $\{\nu \in \mathcal{M}_1^+(TM); \overline{\mathcal{G}}(\nu) \leq \widehat{\mathcal{C}}(\lambda)\}$ is compact. However $\widehat{\mathcal{C}}(\nu) < \infty$ by Lemma 4.1. Since l is super linear in v uniformly in x (see section 1.2-7) it follows that the sub-level set $\{\nu \in \mathcal{M}_\lambda; \int_{TM} l(x, v) d\nu \leq C < \infty\}$ is tight for any constant C , hence compact.

Next

$$\begin{aligned} & \int_{TM} (l(x, v) - \langle d\phi, v \rangle) d\nu(x, v) + \int_M \phi d\lambda \\ &= \int_M \phi d\lambda - \int_M h(x, d\phi) d\mu + \int_{TM} (l(x, v) - \langle d\phi, v \rangle + h(x, d\phi)) d\nu(x, v). \end{aligned} \quad (4.25)$$

where $\mu = \Pi_{\#}\nu$. By the Young inequality $l(x, v) + h(x, \xi) \geq \langle \xi, v \rangle_{(x)}$ for any $\xi \in T_x^*M$, $v \in T_xM$ with equality if and only if $v = h_{\xi}(x, d\phi(x))$. So, the second term on the right of (4.25) is non-negative, but, for any $\mu \in \mathcal{M}_1^+$

$$\inf_{\nu} \left\{ \int_{TM} (l(x, v) - \langle d\phi, v \rangle) d\nu(x, v) \ ; \ \nu \in \mathcal{M}_1^+(TM) , \Pi_{\#}\nu = \mu \right\} = - \int_M h(x, d\phi) d\mu$$

is realized for $\nu = \delta_{v-h_{\xi}(x, d\phi(x))} \oplus \mu \in \mathcal{M}_1^+(TM)$. From this and (4.25) we obtain

$$\inf_{\nu \in \mathcal{M}_1^+(TM)} \mathcal{G}(\nu, \phi) = \inf_{\mu \in \mathcal{M}_1^+} \mathcal{F}(\phi, \mu)$$

hence

$$\sup_{\phi \in C^1(M)} \inf_{\nu \in \mathcal{M}_1^+(TM)} \mathcal{G}(\nu, \phi) = \sup_{\phi \in C^1(M)} \inf_{\mu \in \mathcal{M}_1^+} \mathcal{F}(\phi, \mu) = \widehat{\mathcal{C}}(\lambda)$$

and this part of the Theorem follows from (4.23).

□

4.4 Proof of Theorem 1:(2 \Leftrightarrow 3)

We now define, for any $\lambda \in \mathcal{M}_0$, a measure $\mu_{\lambda} \in \mathcal{M}_1^+$ in the following way:

Assume, for now, that $\lambda \in \mathcal{M}(A)$. If $E(\lambda, T) \in]\underline{E}, \infty[-N$ then define $\mu_{\lambda} = \mu_{\lambda}^{E(\lambda, T)}$ according to Definition 4.1. Otherwise, fix a sequence $E^n \in]\underline{E}, \infty[-N$ such that $E^n \searrow E(\lambda, T)$. Similarly, let $E_n \in]\underline{E}, \infty[-N$ such that $E_n \nearrow E(\lambda, T)$.

Then $\mu_{\lambda_n}^{E_n}$ and $\mu_{\lambda_n}^{E^n}$ are given by Definition 4.1 for any n . Let μ_{λ}^+ be a weak limit of the sequence $\mu_{\lambda_n}^{E_n}$, and, similarly, μ_{λ}^- be a weak limit of the sequence $\mu_{\lambda_n}^{E^n}$.

By Lemma 4.8 and Remark 4.1 we get

$$\int_M d\mu_{\lambda}^+ \leq T \leq \int_M d\mu_{\lambda}^- . \quad (4.26)$$

If $E(\lambda, T) = \underline{E}$ then we can still define μ_{λ}^+ , and it satisfies the left inequality of (4.26).

Definition 4.3. For any $\lambda \in \mathcal{M}_0$, let μ_{λ} defined in the following way:

i) If $\lambda \in \mathcal{M}_0(A)$ then

- If $E(\lambda, T) > \underline{E}$ then μ_{λ} is a convex combination of $T^{-1}\mu_{\lambda}^+, T^{-1}\mu_{\lambda}^-$ given by (4.26) such that $\mu_{\lambda} \in \mathcal{M}_1^+$ (that is, $\int d\mu_{\lambda} = 1$).
- If $E(\lambda, T) = \underline{E}$ then

$$\mu_{\lambda} = T^{-1}\mu_{\lambda}^+ + \left(1 - T^{-1} \int_M d\mu_{\lambda}^+\right) \mu_M \quad (4.27)$$

where μ_M is a projected Mather measure.

ii) For $\lambda \notin \mathcal{M}_0(A)$, let $\lambda_n \in \mathcal{M}_0(A)$ be a sequence converging weakly to λ . Then $\{\mu_{\lambda}\}$ is the set of weak limits of the sequence μ_{λ_n} .

Define

$$\mathcal{Q}(\lambda, \mu) := \sup_{\phi \in C^1(M)} \left\{ - \int_M h(x, d\phi) d\mu + \int_M \phi d\lambda \right\} \in \mathbb{R} \cup \{\infty\}, \quad \mathcal{Q}_T(\lambda, \mu) := \mathcal{Q}(\lambda, T\mu). \quad (4.28)$$

Recall from 1 \Leftrightarrow 2 that

$$\widehat{\mathcal{C}}_T(\lambda) = \inf_{\mu \in \mathcal{M}_1^+} \mathcal{Q}_T(\lambda, \mu) \equiv \inf_{\mu \in \mathcal{M}_1^+} \mathcal{Q}(\lambda, T\mu). \quad (4.29)$$

Also, from (4.13), (4.10) and Proposition 4.1

$$\overline{H}_T^*(\lambda) \leq \mathcal{Q}_T(\lambda, \mu) \quad \forall \mu \in \mathcal{M}_1^+. \quad (4.30)$$

We have to show that

$$\overline{H}_T^*(\lambda) = \inf_{\mu \in \mathcal{M}_1^+} \mathcal{Q}_T(\lambda, \mu) \quad (4.31)$$

for any $\lambda \in \mathcal{M}_0$. It is enough to prove (4.31) for a dense set in \mathcal{M}_0 , say for any $\lambda \in \mathcal{M}_0(A)$. Suppose (4.31) holds for a sequence $\{\lambda_n\} \subset \mathcal{M}_0(A)$ converging weakly to $\lambda \in \mathcal{M}_0$, that is, $\overline{H}_T^*(\lambda_n) = \widehat{\mathcal{C}}_T(\lambda_n)$. Since \overline{H}_T^* is weakly continuous by Corollary 4.3 we get $\overline{H}_T^*(\lambda) = \lim_{n \rightarrow \infty} \overline{H}_T^*(\lambda_n)$. On the other hand we recall that, according to definition 2 of Theorem 1, $\widehat{\mathcal{C}}_T : \mathcal{M}_0 \mapsto \mathbb{R}$ is l.s.c. So $\lim_{n \rightarrow \infty} \widehat{\mathcal{C}}_T(\lambda_n) \geq \widehat{\mathcal{C}}_T(\lambda)$, hence $\overline{H}_T^*(\lambda) \geq \widehat{\mathcal{C}}_T(\lambda)$. By (4.29, 4.30) we get (4.31) for any $\lambda \in \mathcal{M}_0$.

The proof of 2 \Leftrightarrow 3 then follows from

Lemma 4.9. *For any $\lambda \in \mathcal{M}_0(A)$*

$$\mathcal{Q}_T(\lambda, \mu_\lambda) = \overline{H}_T^*(\lambda) \quad (4.32)$$

holds where $\mu_\lambda \in \mathcal{M}_1^+$ is as given in Definition 4.3.

Proof. Let $\lambda \in \mathcal{M}_0(A)$ and $E \in]\underline{E}, \infty[-N$. Then we use (4.21) for any $\phi \in C^1(M)$

$$- \int_M h(x, d\phi) d\mu_\lambda^E = - \sum_{x,y \in A} \Lambda(\{x, y\}) \int_0^1 h(z_{x,y}^E(s), d\phi(z_{x,y}^E(s))) ds$$

We now perform a change of variables $ds \rightarrow dt = \sigma'_E(z_{x,y}^E(s), \dot{z}_{x,y}^E(s)) ds$ which transforms the interval $[0, 1]$ into $[0, T_E(x, y)]$ (see (4.7)) and we get

$$- \int_M h(x, d\phi) d\mu_\lambda^E = - \sum_{x,y \in A} \Lambda(\{x, y\}) \int_0^{T_E(x,y)} h(\widehat{z}_{x,y}^E(t), d\phi(\widehat{z}_{x,y}^E(t))) dt$$

where $\widehat{z}_{x,y}^E$ is the re-parametrization of $z_{x,y}^E$, satisfying $\widehat{z}_{x,y}^E(0) = x$, $\widehat{z}_{x,y}^E(T_E(x, y)) = y$. Next

$$\int_M \phi d\lambda = \int_M d\Lambda_\lambda^E(x, y) [\phi(y) - \phi(x)] = \sum_{x,y \in A} \Lambda(\{x, y\}) \int_0^{T_E(x,y)} d\phi(\widehat{z}_{x,y}^E(t)) \dot{\widehat{z}}_{x,y}^E(t) dt$$

so $\int_M \phi d\lambda - \int_M h(x, d\phi) d\mu_\lambda^E =$

$$\begin{aligned}
& \sum_{x,y \in A} \Lambda_\lambda^E(\{x, y\}) \int_0^{T_E(x,y)} \left[d\phi \left(\widehat{z}_{x,y}^E(t) \right) \dot{\widehat{z}}_{x,y}^E(t) - h \left(\widehat{z}_{x,y}^E(t), d\phi \left(\widehat{z}_{x,y}^E(t) \right) \right) \right] dt \\
& \leq \sum_{x,y \in A} \Lambda_\lambda^E(\{x, y\}) \int_0^{T_E(x,y)} l \left(\widehat{z}_{x,y}^E(t), \dot{\widehat{z}}_{x,y}^E(t) \right) dt = \sum_{x,y \in A} \Lambda_\lambda^E(\{x, y\}) C_{T_E(x,y)}(x, y) \\
& = \sum_{x,y \in A} \Lambda_\lambda^E(\{x, y\}) [C_{T_E(x,y)}(x, y) + ET_E(x, y)] - E \sum_{x,y \in A} \Lambda_\lambda^E(\{x, y\}) T_E(x, y) = \\
& \sum_{x,y \in A} \Lambda_\lambda^E(\{x, y\}) D_E(x, y) - E \sum_{x,y \in A} \Lambda_\lambda^E(\{x, y\}) T_E(x, y) = \mathcal{D}_E(\lambda) - ED'_E(\lambda). \quad (4.33)
\end{aligned}$$

To obtain (4.33) we used the Young inequality in the second line, (4.8) and (4.17) on the last line.

Since (4.33) is valid for any $\phi \in C^1(M)$ we get from this and (4.30) that

$$\mathcal{D}_E(\lambda) - ED'_E(\lambda) \geq \mathcal{Q}(\lambda, \mu_\lambda^E) \geq \overline{H}_T^*(\lambda) = \max_{E \geq \underline{E}} \mathcal{D}_E(\lambda) - TE, \quad (4.34)$$

holds for *any* $E \geq \underline{E}$. Now, if it so happens that the maximizer $E(\lambda, T)$ on the right of (4.34) is on the complement of the set N in $[\underline{E}, \infty[$, then $D'_E(\lambda) = T = \int_M d\mu_\lambda^E$ for $E = E(\lambda, T)$ via Lemma 4.8 and the inequality in (4.34) turns into an equality. Otherwise, if $E(\lambda, T) \in N - \{\underline{E}\}$, we take the sequences $E_n \nearrow E(\lambda, T)$, $E^n \searrow E(\lambda, T)$ for $E_n, E^n \in]\underline{E}, \infty[-N$ and the corresponding limits μ_λ^+ , μ_λ^- defined in (4.26). Since \mathcal{Q}_T is a convex, l.s.c as a function of μ we get that the left inequality in (4.34) survives the limit, and

$$\mathcal{D}_{E(\lambda, T)}(\lambda) - E(\lambda, T) \frac{d^+}{dE} \mathcal{D}_{E(\lambda, T)}(\lambda) \geq \mathcal{Q}(\lambda, \mu_\lambda^+), \quad \mathcal{D}_{E(\lambda, T)}(\lambda) - E(\lambda, T) \frac{d^-}{dE} \mathcal{D}_{E(\lambda, T)}(\lambda) \geq \mathcal{Q}(\lambda, \mu_\lambda^-), \quad (4.35)$$

while $\frac{d^+}{dE} \mathcal{D}_{E(\lambda, T)}(\lambda) = \int d\mu_\lambda^+$ and $\frac{d^-}{dE} \mathcal{D}_{E(\lambda, T)}(\lambda) = \int d\mu_\lambda^-$. Then, upon taking a convex combination $\mu_\lambda = \alpha T^{-1} \mu_\lambda^+ + T^{-1} (1 - \alpha) \mu_\lambda^-$ such that, according to Definition 4.3,

$$\alpha \frac{d^+}{dE} \mathcal{D}_{E(\lambda, T)}(\lambda) + (1 - \alpha) \frac{d^-}{dE} \mathcal{D}_{E(\lambda, T)}(\lambda) = T \int d\mu_\lambda = T \quad (4.36)$$

and using the convexity of \mathcal{Q} in μ we get from (4.35, 4.36)

$$\mathcal{D}_{E(\lambda, T)}(\lambda) - TE(\lambda, T) \geq \mathcal{Q}(\lambda, T\mu_\lambda) \equiv \mathcal{Q}_T(\lambda, \mu_\lambda)$$

This, with the right inequality of (4.32) yields the equality $\mathcal{Q}_T(\lambda, \mu_\lambda) = \overline{H}_T^*(\lambda)$.

Finally, if $E(\lambda, T) = \underline{E}$ we proceed as follows: Let $E^n \searrow \underline{E}$ and $\mu_\lambda^+ := \lim_{n \rightarrow \infty} \mu_\lambda^{E^n}$. It follows that

$$\int_M d\mu_\lambda^+ = \lim_{n \rightarrow \infty} \int_M d\mu_\lambda^{E^n} = \lim_{n \rightarrow \infty} \mathcal{D}'_{E^n}(\lambda) = \frac{d^+}{dE} \mathcal{D}_{\underline{E}}(\lambda) \in (0, T]. \quad (4.37)$$

Let μ_λ as in (4.27). From (4.28, , 4.37) and (2.4) we get

$$\mathcal{Q}_T(\lambda, \mu_\lambda) \leq \mathcal{Q}(\lambda, \mu_\lambda^+) + \left(T - \frac{d^+}{dE} \mathcal{D}_{\underline{E}}(\lambda) \right) \mathcal{Q}(0, \mu_M) = \mathcal{Q}(\lambda, \mu_\lambda^+) - \left(T - \frac{d^+}{dE} \mathcal{D}_{\underline{E}}(\lambda) \right) \underline{E} \quad (4.38)$$

while (2.4) and the left part of (4.35) for $E = \underline{E}$ imply

$$\mathcal{Q}(\lambda, \mu_\lambda^+) \leq \mathcal{D}_{\underline{E}}(\lambda) - \underline{E} \frac{d^+}{dE} \mathcal{D}_{\underline{E}}(\lambda) . \quad (4.39)$$

From (4.38) and (4.39) we get

$$\mathcal{Q}_T(\lambda, \mu_\lambda) \leq \mathcal{D}_{\underline{E}}(\lambda) - \underline{E} T \leq \overline{H}_T^*(\lambda)$$

and the equality holds via (4.30). The last part of Theorem 1 follows from the equality in (4.30) as well. \square

4.5 Proof of Theorem 3

Theorem 1-(2) and (3.6) imply

$$\widehat{\mathcal{C}}_T(\lambda) = \min_{\mu \in \mathcal{M}_1^+} \widehat{\mathcal{C}}_T(\lambda \| \mu) . \quad (4.40)$$

Next, we note that $\mathcal{D}_E(\lambda \| \mu)$ is a concave function of E for $E \geq \underline{E}$. In fact, from (3.4) and convexity of $h(x, \cdot)$ for each $x \in M$ we obtain

$$\phi_i \in \mathcal{H}_{E_i} , i = 1, 2 \implies \alpha \phi_1 + (1 - \alpha) \phi_2 \in \mathcal{H}_{\alpha E_1 + (1 - \alpha) E_2}$$

for $\alpha \in (0, 1)$ and $E_1, E_2 \geq \underline{E}$. The concavity of $\mathcal{D}_{(\cdot)}(\lambda \| \mu)$ follows from its definition (3.5). Then, by convex duality and (3.6)

$$\mathcal{D}_E(\lambda \| \mu) = \min_{T > 0} \left[\widehat{\mathcal{C}}_T(\lambda \| \mu) + ET \right] .$$

By the same argument

$$\mathcal{D}_E(\lambda) = \min_{T > 0} \left[\widehat{\mathcal{C}}_T(\lambda) + ET \right] .$$

Hence, (4.40) and Theorem 1-(3) imply

$$\begin{aligned} \min_{\mu \in \mathcal{M}_1^+} \mathcal{D}_E(\lambda \| \mu) &= \min_{\mu \in \mathcal{M}_1^+} \min_{T > 0} \left[\widehat{\mathcal{C}}_T(\lambda \| \mu) + ET \right] \\ &= \min_{T > 0} \min_{\mu \in \mathcal{M}_1^+} \left[\widehat{\mathcal{C}}_T(\lambda \| \mu) + ET \right] = \min_{T > 0} \left[\widehat{\mathcal{C}}_T(\lambda) + ET \right] = \mathcal{D}_E(\lambda) . \end{aligned}$$

\square

5 Proof of Theorems 2&4

5.1 Auxiliary results

Lemma 5.1 follows from the surjectivity of $Exp_l^{(t)}(x)$ as a mapping from $T_x M$ to M , for any $x \in M$ and any $t \neq 0$ (Recall definition at Section 1.2-8):

Lemma 5.1. *Let $\Lambda \in \mathcal{M}^+(M \times M)$. For any $t > 0$ there exists a Borel measure $\widehat{\Lambda}^{(t)} \in \mathcal{M}^+(TM)$ such that $(I \otimes Exp_l^{(t)})_{\#} \widehat{\Lambda}^{(t)} = \Lambda$. Here $I \otimes Exp_l^{(t)}(x, v) := (x, Exp_l^{(t)}(x, v))$.*

The proof of Lemma 5.2 follows directly from the definition of the optimal plan:

Lemma 5.2. *Let Λ be a minimizer for (2.6), $B \subset M \times M$ a Borel subset and $\Lambda|_B$ the restriction of Λ to B . Let μ_B^0, μ_B^1 the marginals of $\Lambda|_B$ on the factors of $M \times M$. Then $\Lambda|_B$ is an optimal plan for $\mathcal{C}(\mu_B^0, \mu_B^1)$. In addition, if $B_1, B_2 \subset M \times M$ are disjoint Borel sets then*

$$\mathcal{C}(\mu_{B_1}^0, \mu_{B_1}^1) + \mathcal{C}(\mu_{B_2}^0, \mu_{B_2}^1) = \mathcal{C}(\mu_{B_1}^0 + \mu_{B_2}^0, \mu_{B_1}^1 + \mu_{B_2}^1)$$

and $\Lambda|_{B_1 \cup B_2}$ is the optimal plan with respect to $\mathcal{C}(\mu_{B_1}^0 + \mu_{B_2}^0, \mu_{B_1}^1 + \mu_{B_2}^1)$.

Lemma 5.3 represents the *time interpolation* of optimal plans (see [28]):

Lemma 5.3. *Given $t > 0$ and $\lambda = \lambda^+ - \lambda^- \in \mathcal{M}_0$. Let $\Lambda^t \in \mathcal{P}(\lambda^+, \lambda^-)$ be an optimal plan realizing*

$$\mathcal{C}_t(\lambda^+, \lambda^-) = \int \int C_t(x, y) \Lambda^t(dx dy) .$$

Let $\widehat{\Lambda}^{(t)} \in \mathcal{M}^+(TM)$ given in Lemma 5.1 for $\Lambda = \Lambda^t$. Let $\lambda_s := (Exp_l^{(s)})_{\#} \widehat{\Lambda}^{(t)}$. Then, if $0 < s < t$,

$$\mathcal{C}_s(\lambda^+, \lambda_s) + \mathcal{C}_{t-s}(\lambda_s, \lambda^-) = \mathcal{C}_t(\lambda^+, \lambda^-) .$$

Lemma 5.4. *For any $\lambda^+, \lambda^- \in \mathcal{M}_1^+$ satisfying $\lambda = \lambda^+ - \lambda^- \in \mathcal{M}_1^+$,*

$$\mathcal{C}_T(\lambda^+, \lambda^-) \geq \widehat{\mathcal{C}}_T(\lambda) .$$

Lemma 5.5. *$\widehat{\mathcal{C}}_T(\lambda|\mu)$ is l.s.c in the weak-* topology of $\mathcal{M}_0 \times \mathcal{M}_1^+$. Assuming \mathbf{H}_1 and \mathbf{H}_2 , for any $\lambda \in \mathcal{M}_0$, $\mu \in \mathcal{M}_1^+$ there exists a sequence $\{\tilde{\mu}_n\} = \{\rho_n(x)dx\} \subset \mathcal{M}_1^+$, $\{\tilde{\lambda}_n\} = \{\rho_n(q_n^+ - q_n^-)dx\} \subset \mathcal{M}_0$ where $\rho_n \in C^\infty(M)$ are positive everywhere, $q_n^\pm \in C^\infty(M)$ non-negatives such that $\tilde{\lambda}_n \rightarrow \lambda$, $\tilde{\mu}_n \rightarrow \mu$ and*

$$\lim_{n \rightarrow \infty} \widehat{\mathcal{C}}_T(\tilde{\lambda}_n|\tilde{\mu}_n) = \widehat{\mathcal{C}}_T(\lambda|\mu) . \quad (5.1)$$

Lemma 5.6. *For any $\mu \in \mathcal{M}_1^+$, $\lambda = \lambda^+ - \lambda^- \in \mathcal{M}_0$*

$$\liminf_{\varepsilon \rightarrow 0} \varepsilon^{-1} \mathcal{C}_{\varepsilon T}(\mu + \varepsilon \lambda^+, \mu + \varepsilon \lambda^-) \geq \widehat{\mathcal{C}}_T(\lambda|\mu) .$$

Lemma 5.7. *Assume $\mu = \rho(x)dx$ and $\lambda = \rho(q^+ - q^-)dx$ where ρ, q^\pm are C^∞ functions, ρ positive everywhere on M . Then*

$$\limsup_{\varepsilon \rightarrow 0} \varepsilon^{-1} \mathcal{C}_{\varepsilon T}(\mu + \varepsilon \lambda^+, \mu + \varepsilon \lambda^-) \leq \widehat{\mathcal{C}}_T(\lambda|\mu) .$$

Lemma 5.8. For $T > 0$,

$$\widehat{\mathcal{C}}_T(\lambda) \geq \limsup_{\varepsilon \rightarrow 0} \varepsilon^{-1} \inf_{\mu \in \mathcal{M}_1^+} \mathcal{C}_{\varepsilon T}(\mu + \varepsilon\lambda^+, \mu + \varepsilon\lambda^-) .$$

Proof. of Lemma 5.4: We use the duality representation of the Monge-Kantorovich functional [26] to obtain (recall $\lambda^\pm \in \mathcal{M}_1^+$)

$$\mathcal{C}_T(\lambda^+, \lambda^-) + ET = \sup_{\psi, \phi} \left\{ \int_M \psi d\lambda^- - \phi d\lambda^+ \quad , \quad \phi(y) - \psi(x) \leq C_T(x, y) + ET \right\}$$

By (2.10) $C_T(x, y) + ET \geq D_E(x, y)$ for any $x, y \in M$ so, by (2.12, 2.13)

$$\begin{aligned} \sup_{\psi, \phi} \left\{ \int_M \psi d\lambda^- - \phi d\lambda^+ \quad , \quad \phi(y) - \psi(x) \leq C_T(x, y) + ET \right\} &\geq \sup_{\phi} \left\{ \int_M \phi d\lambda \quad , \quad \phi(y) - \phi(x) \leq D_E(x, y) \right\} \\ &= \mathcal{D}_E(\lambda) \quad (5.2) \end{aligned}$$

so

$$\mathcal{C}_T(\lambda^+, \lambda^-) \geq \mathcal{D}_E(\lambda) - ET$$

for any $E \geq \underline{E}$. By Theorem 1-(3)

$$\mathcal{C}_T(\lambda^+, \lambda^-) \geq \sup_{E \geq \underline{E}} \mathcal{D}_E(\lambda) - ET = \widehat{\mathcal{C}}_T(\lambda) .$$

□

Proof. of Lemma 5.5: From (3.5, 3.6) we obtain

$$\widehat{\mathcal{C}}_T(\lambda \parallel \mu) = \sup_{\phi \in C^1(M)} \int_M \phi d\lambda - Th(x, d\phi) d\mu .$$

In particular $\widehat{\mathcal{C}}_T$ is l.s.c (and convex) on $\mathcal{M}_0 \times \mathcal{M}_1^+$.

Let $\varepsilon_n \rightarrow 0$ and $\lambda_n := \lambda_{\varepsilon_n} := \delta_{\varepsilon_n} * \lambda \in \mathcal{M}_0$ defined by

$$\int_M \psi d\lambda_n := \lambda(\delta_{\varepsilon_n} * \psi) \quad \forall \psi \in C^0(M) . \quad (5.3)$$

By **H₁**, $\lambda_n \rightharpoonup \lambda$ while λ_n are smooth. First, we observe that $\lim_{n \rightarrow \infty} \lambda_n \rightharpoonup \lambda$. Indeed, for any $\psi \in C^1(M)$:

$$\lim_{n \rightarrow \infty} \int_M \psi d\lambda_n = \lim_{n \rightarrow \infty} \lambda(\delta_{\varepsilon_n} * \psi) = \lambda(\psi) .$$

Next, by Jensen's Theorem and **H₂**

$$\begin{aligned} \int_M h(x, d\delta_\varepsilon * \phi) d\mu &= \int_M h(x, \delta_\varepsilon * d\phi) d\mu \leq \int_{M \times M} h(x, d\phi(y)) \delta_\varepsilon(x, y) d\mu(x) dy \\ &\equiv \int_M h(x, d\phi) d\delta_\varepsilon * \mu + \int_{M \times M} [h(x, d\phi(y)) - h(y, d\phi(y))] \delta_\varepsilon(x, y) d\mu(x) dy \quad (5.4) \end{aligned}$$

From section 1.2-(7) and using $\delta_\varepsilon(x, y) = o(1)$ for $D(x, y) > \delta$,

$$\int_{M \times M} [h(x, d\phi(y)) - h(y, d\phi(y))] \delta_\varepsilon(x, y) d\mu(x) dy \leq O(\varepsilon) + o(1) \int_M h(x, d\phi) d\delta_\varepsilon * \mu .$$

Next, define $\mu_n = \delta_{\varepsilon_n} * \mu$. Let ψ_n be the maximizer of $\widehat{\mathcal{C}}(\lambda_n \| \mu_n)$, that is

$$\widehat{\mathcal{C}}_T(\lambda_n \| \mu_n) = \int_M \psi_n d\lambda_n - Th(x, d\psi_n) d\mu_n$$

By (5.3, 5.4)

$$\begin{aligned} \widehat{\mathcal{C}}_T(\lambda_n \| \mu_n) &\leq \int_M \delta_\varepsilon * \psi_n d\lambda - (1 - o(1)) \int_M Th(x, d\delta_\varepsilon * \psi_n) d\mu + O(\varepsilon_n) = \\ (1 - o(1)) &\left[\int_M \delta_\varepsilon * \psi_n \frac{d\lambda}{1 - o(1)} - \int_M Th(x, d\delta_\varepsilon * \psi_n) d\mu \right] + \varepsilon_n \leq (1 - o(1)) \widehat{\mathcal{C}} \left(\frac{\lambda}{1 - o(1)} \| \mu \right) + \varepsilon_n \end{aligned} \quad (5.5)$$

We obtained

$$\limsup_{n \rightarrow \infty} \widehat{\mathcal{C}}_T(\lambda_n \| \mu_n) \leq \widehat{\mathcal{C}}_T(\lambda \| \mu)$$

which, together with the l.s.c of $\widehat{\mathcal{C}}_T$, implies the result. \square

Proof. of Lemma 5.6: Recall that the Lax-Oleinik Semigroup acting on $\phi \in C^0(M)$

$$\psi(x, t) = LO(\phi)_{(t, x)} := \sup_{y \in M} [\phi(y) - C_t(x, y)]$$

is a viscosity solution of the Hamilton-Jacobi equation $\partial_t \psi - h(x, d\psi) = 0$ subjected to $\psi_0 = \phi(x)$. If $\phi \in C^1(M)$ then ψ is a *classical solution* on some neighborhood of $t = 0$, so

$$\lim_{T \rightarrow 0} LO(\phi)_{(T, \cdot)} = \phi \quad ; \quad \lim_{T \rightarrow 0} T^{-1} [LO(\phi)_{(T, x)} - \phi(x)] = h(x, d\phi) .$$

Then for any $\mu_1, \mu_2 \in \mathcal{M}_1^+$

$$\begin{aligned} \mathcal{C}_T(\mu_1, \mu_2) &= \sup_{\phi, \psi \in C^1(M)} \left\{ \int_M \phi d\mu_2 - \psi d\mu_1 \quad ; \quad \phi(x) - \psi(y) \leq C_T(x, y) \quad \forall x, y \in M \right\} = \\ &\sup_{\phi \in C^1(M)} \int_M \phi d\mu_2 - LO(\phi)_{(T, x)} d\mu_1 \end{aligned} \quad (5.6)$$

Hence

$$\begin{aligned} \liminf_{\varepsilon \rightarrow 0} \varepsilon^{-1} \mathcal{C}_{\varepsilon T}(\mu + \varepsilon\lambda^+, \mu + \varepsilon\lambda^-) &= \\ \liminf_{\varepsilon \rightarrow 0} \sup_{\phi \in C^1(M)} \int_M \varepsilon^{-1} [\phi(x) - LO(\phi)_{(\varepsilon T, x)}] d\mu + \int_M \phi d\lambda^+ - LO(\phi)_{(\varepsilon T, x)} d\lambda^- & \\ \geq \sup_{\phi \in C^1(M)} \lim_{\varepsilon \rightarrow 0} \int_M \varepsilon^{-1} [\phi(x) - LO(\phi)_{(\varepsilon T, x)}] d\mu + \int_M \phi d\lambda^+ - LO(\phi)_{(\varepsilon T, x)} d\lambda^- & \\ = \sup_{\phi, \psi \in C^1(M)} \int_M -Th(x, d\phi) d\mu + \phi d\lambda := \widehat{\mathcal{C}}_T(\lambda \| \mu) . \end{aligned} \quad (5.7)$$

\square

Proof. of Lemma 5.7: We may describe the optimal mapping $S_{\varepsilon T} : M \rightarrow M$ associated with $C_{\varepsilon T}(\mu + \varepsilon\lambda^+, \mu + \varepsilon\lambda^-)$ in local coordinates on each chart. It is given by the solution to the Monge-Ampère equation

$$\det \nabla_x S_{\varepsilon T} = \frac{\rho(x)(1 + \varepsilon q^-(x))}{\rho(S_{\varepsilon T}(x))(1 + \varepsilon T q^+(S_{\varepsilon T}(x)))} \quad (5.8)$$

where

$$\nabla \psi = -\nabla_x C_{\varepsilon T}(x, S_{\varepsilon T}(x)) \quad (5.9)$$

and

$$C_{\varepsilon T}(\mu + \varepsilon\lambda^+, \mu + \varepsilon\lambda^-) = \int_M C_{\varepsilon T}(x, S_{\varepsilon T}(x)) \rho(1 + \varepsilon T q^-) dx \quad (5.10)$$

We recall that the inverse of $\nabla_x C_{\varepsilon T}(x, \cdot)$ with respect to the second variable is $I_d + \varepsilon T \nabla \psi$, to leading order in ε . That is,

$$\nabla_x C_{\varepsilon T}(x, x + \varepsilon T \partial_\rho h(x, \xi) + (\varepsilon T)^2 Q(x, \xi, \varepsilon)) = -\xi \quad (5.11)$$

where (here and below) Q is a generic smooth function of its arguments.

Hence, $S_{\varepsilon T}$ can be expanded in ε in terms of ψ as

$$S_{\varepsilon T}(x) = x + \varepsilon T h_\xi(x, \nabla \psi) + (\varepsilon T)^2 Q(x, \nabla \psi, \varepsilon) \quad (5.12)$$

We now expand the right side of (5.8) using (5.12) to obtain

$$1 + \varepsilon T [q^-(x) - q^+(x) - h_\xi(x, d\psi) \cdot \nabla_x \ln \rho(x)] + (\varepsilon T)^2 Q(x, \nabla \psi, x, \varepsilon) \quad (5.13)$$

while the left hand side is

$$\det(\nabla_x S_{\varepsilon T}) = 1 + \varepsilon T \nabla \cdot h_\xi(x, d\psi) + (\varepsilon T)^2 Q(x, \nabla \psi, \nabla \nabla \psi, x, \varepsilon) \quad (5.14)$$

Comparing (5.13, 5.14), divide by εT and multiply by ρ to obtain

$$T \nabla \cdot (\rho h_\xi(x, d\psi)) = \rho(q^- - q^+) + \varepsilon T \rho Q(x, \nabla \psi, \nabla \nabla \psi, x, \varepsilon). \quad (5.15)$$

Now, we substitute $\varepsilon = 0$ and get a quasi-linear equation for ψ_0 :

$$T \nabla \cdot (\rho h_\xi(x, d\psi_0)) = \rho(q^- - q^+). \quad (5.16)$$

ψ_0 is a maximizer of

$$\widehat{\mathcal{C}}_T(\lambda \parallel \mu) = \int_M \rho(q^+ - q^-) \psi_0 - \int_M \rho T h(x, d\psi_0) dx$$

By elliptic regularity, $\psi_0 \in C^\infty(M)$. Multiply (5.16) by ψ_0 and integrate over M to obtain

$$\int_M \rho(q^+ - q^-) = \int_M \rho T h_\xi(x, d\psi_0) \cdot \nabla \psi_0$$

Then by the Lagrangian/Hamiltonian duality

$$\widehat{\mathcal{C}}_T(\lambda\|\mu) = \int_M \rho T [\nabla\psi_0 \cdot h_\xi(x, d\psi_0) - h(x, d\psi_0)] \equiv T \int_M \rho l(x, h_\xi(x, d\psi_0)) . \quad (5.17)$$

We observe $l(x, \frac{y-x}{T}) \geq T^{-1}C_T(x, y)$. So, (5.10) with (5.12) imply

$$(\varepsilon T)^{-1}C_{\varepsilon T}(\mu + \varepsilon\lambda^+, \mu + \varepsilon\lambda^-) \leq \int_M \rho(1 + \varepsilon T q^-)l(x, h_\xi(x, \nabla\psi_\varepsilon + \varepsilon T Q(x, \nabla\psi_\varepsilon, \varepsilon))) \quad (5.18)$$

where ψ_ε is a solution of (5.15). Now, if we show that $\lim_{\varepsilon \rightarrow 0} \psi_\varepsilon = \psi_0$ in $C^1(M)$ then, from (5.17, 5.18)

$$\limsup_{\varepsilon \rightarrow 0} (\varepsilon)^{-1}C_{\varepsilon T}(\mu + \varepsilon\lambda^+, \mu + \varepsilon\lambda^-) \leq T \int_M \rho l(x, h_\xi(x, d\psi_0)) = \widehat{\mathcal{C}}(\lambda\|\mu) .$$

Next we show that, indeed, $\lim_{\varepsilon \rightarrow 0} \psi_\varepsilon = \psi_0$ in $C^1(M)$.

Substitute $\psi_\varepsilon = \psi_0 + \phi_\varepsilon$ in (5.15). We obtain

$$\nabla \cdot (\sigma(x)\nabla\phi_\varepsilon) = \varepsilon Q(x, \nabla\phi_\varepsilon, \nabla\nabla\phi_\varepsilon, \varepsilon) + \nabla \cdot (\rho\langle \nabla^t\phi_\varepsilon, \tilde{Q}(x, \nabla\phi, \varepsilon) \cdot \nabla\phi_\varepsilon \rangle) \quad (5.19)$$

where $\sigma := Th_{\xi\xi}(x, \nabla\psi_0(x))$ is a positive definite form, while \tilde{Q} is a smooth matrix valued functions in both x and ε , determined by $\nabla\psi_0$ and Q as given in (5.15). A direct application of the implicit function theorem implies the existence of a branch $(\lambda(\varepsilon), \eta_\varepsilon)$ of solutions for

$$\nabla \cdot (\sigma(x)\nabla\eta) = \varepsilon Q(x, \nabla\eta, \nabla\nabla\eta, \varepsilon) + \nabla \cdot (\rho\langle \nabla^t\eta, \tilde{Q}(x, \nabla\eta, \varepsilon) \circ \nabla\eta \rangle) + \lambda(\varepsilon) \quad (5.20)$$

where $\eta_0 = \lambda(0) = 0$ and $\varepsilon \mapsto \eta_\varepsilon$ is (at least) continuous in $C^1(M) \perp 1$. Note that for $\varepsilon \neq 0$ we may have a non-zero $\lambda(\varepsilon)$ which follows from projecting the right side on the equation to the Hilbert space perpendicular to constants (recall that M is a compact manifold without boundary, and the left side is surjective on this space). We now show that $\eta_\varepsilon = \phi_\varepsilon$, i.e $\lambda(\varepsilon) = 0$ also for $\varepsilon \neq 0$. Indeed, (5.19) is equivalent to (5.8) multiplied by $\rho(x)/\varepsilon$, so (5.20) is equivalent to

$$\det\nabla_x \hat{S}_{\varepsilon T} = \frac{\rho(x)(1 + \varepsilon q^-(x))}{\rho(\hat{S}_{\varepsilon T}(x))(1 + \varepsilon q^+(\hat{S}_{\varepsilon T}(x)))} + \varepsilon \rho^{-1}(x)\lambda(\varepsilon)$$

where $\hat{S}_{\varepsilon T}(x)$ obtained from (5.12) with $\psi_\varepsilon := \psi_0 + \eta_\varepsilon$.

Hence

$$\begin{aligned} \int_M (\rho(\hat{S}_{\varepsilon T}(x))(1 + \varepsilon q^+(\hat{S}_{\varepsilon T}(x)))) \det(\nabla_x \hat{S}_{\varepsilon T}) &= \int_M (\rho(x)(1 + \varepsilon q^-(x))) \\ &+ \varepsilon \lambda(\varepsilon) \int_M \frac{\rho(\hat{S}_{\varepsilon T}(x))}{\rho(x)} (1 + \varepsilon q^+(\hat{S}_{\varepsilon T}(x))) \quad (5.21) \end{aligned}$$

However, $\hat{S}_{\varepsilon T}(x) = x + O(\varepsilon)$ is a diffeomorphism on M , so

$$\begin{aligned} \int_M \left(\rho(\hat{S}_{\varepsilon T}(x))(1 + \varepsilon q^+(\hat{S}_{\varepsilon T}(x))) \right) \det(\nabla_x \hat{S}_{\varepsilon T}) &= \int_M \left(\rho(\hat{S}_{\varepsilon T}(x))(1 + Tq^+(\hat{S}_{\varepsilon T}(x))) \right) |\det(\nabla_x \hat{S}_{\varepsilon T})| \\ &= \int_M \rho(x)(1 + \varepsilon q^+(x)) \equiv \int_M \rho(x)(1 + \varepsilon q^-(x)) . \end{aligned} \quad (5.22)$$

It follows that

$$\varepsilon \lambda(\varepsilon) \int_M \frac{\rho(\hat{S}_{\varepsilon T}(x))}{\rho(x)} (1 + \varepsilon q^+(\hat{S}_{\varepsilon T}(x))) = 0 .$$

Since ρ is positive everywhere it follows that $\lambda(\varepsilon) \equiv 0$ for $|\varepsilon|$ sufficiently small. We proved that $\eta_\varepsilon \equiv \phi_\varepsilon$ and, in particular, $\phi_\varepsilon \rightarrow 0$ as $\varepsilon \rightarrow 0$ in $C^1 \perp 1$, which implies the convergence of ψ_ε to ψ_0 at $\varepsilon \rightarrow 0$ in $C^1 \perp 1$. \square

Proof. (of Lemma 5.8) Given $\varepsilon > 0$ let

$$D_E^\varepsilon(x, y) := \inf_{n \in \mathbb{N}} [C_{\varepsilon n T}(x, y) + \varepsilon n E T] . \quad (5.23)$$

Evidently, $D_E^\varepsilon(x, y)$ is continuous on $M \times M$ locally uniformly in $E \geq \underline{E}$. Moreover,

$$\lim_{\varepsilon \searrow 0} D_E^\varepsilon = D_E \quad (5.24)$$

uniformly on $M \times M$ and locally uniformly in $E \geq \underline{E}$ as well.

We now decompose $M \times M$ into mutually disjoint Borel sets Q_n :

$$M \times M = \cup_n Q_n^\varepsilon , \quad Q_n^\varepsilon \cap Q_{E, n'}^\varepsilon = \emptyset \text{ if } n \neq n'$$

such that

$$Q_n^\varepsilon \subset \{(x, y) \in M \times M ; D_E^\varepsilon(x, y) = C_{\varepsilon n T}(x, y) + \varepsilon n E T\} .$$

Let $\Lambda_\varepsilon^E \in \mathcal{P}(\lambda^+, \lambda^-)$ be an optimal plan for

$$\mathcal{D}_E^\varepsilon(\lambda) = \int_{M \times M} D_E^\varepsilon(x, y) d\Lambda_\varepsilon^E = \min_{\Lambda \in \mathcal{P}(\lambda^+, \lambda^-)} \int_{M \times M} D_E^\varepsilon(x, y) d\Lambda , \quad (5.25)$$

and $\Lambda_\varepsilon^n = \Lambda_\varepsilon^E|_{Q_n^\varepsilon}$, the restriction of Λ_ε^E to Q_n^ε . Set λ_n^\pm to be the marginals of Λ_ε^n on the first and second factors of $M \times M$. Then $\sum_{n=1}^\infty \Lambda_\varepsilon^n = \Lambda_\varepsilon^E$ and

$$\sum_{n=1}^\infty \lambda_n^\pm = \lambda^\pm \quad (5.26)$$

Remark 5.1. Note that $Q_n^\varepsilon = \emptyset$ for all but a finite number of $n \in \mathbb{N}$. In particular, the sum (5.26) contains only a finite number of non-zero terms.

Let $|\lambda_n| := \int_M d\lambda_n^\pm \equiv \int_{M \times M} d\Lambda_n^\varepsilon$. The *averaged flight time* is

$$\langle T \rangle^\varepsilon := \varepsilon T \sum_{n=1}^{\infty} n |\lambda_n| \quad (5.27)$$

We observe that $\langle T \rangle^\varepsilon \in \partial_E \mathcal{D}_E^\varepsilon(\lambda)$, where ∂_E is the super gradient as a function of E . At this stage we choose E depending on ε, T such that

$$\langle T \rangle^\varepsilon = T + 2\varepsilon T |\lambda^\pm| \quad (5.28)$$

We now apply Lemma 5.1: Recalling Section 1.2-8, let $\widehat{\Lambda}_\varepsilon^n \in \mathcal{M}^+(TM)$ satisfying $(I \oplus \text{Exp}_{(l)}^{(t=\varepsilon n T)})_{\#} \widehat{\Lambda}_\varepsilon^n = \Lambda_\varepsilon^n$. Use $\widehat{\Lambda}_\varepsilon^n$ to define $\lambda_n^j := (\text{Exp}_{(l)}^{(t=\varepsilon n T)})_{\#} \widehat{\Lambda}_\varepsilon^n \in \mathcal{M}^+(M)$ for $j = 0, 1 \dots n$. Note that

$$\lambda_n^0 = \lambda_n^+ \quad , \quad \lambda_n^n = \lambda_n^- \quad . \quad (5.29)$$

By Lemma 5.3

$$\mathcal{C}_{\varepsilon n T}(\lambda_n^+, \lambda_n^-) + \varepsilon n E T |\lambda_n| = \sum_{j=0}^{n-1} [\mathcal{C}_{\varepsilon T}(\lambda_n^j, \lambda_n^{j+1}) + \varepsilon E T |\lambda_n|] \quad (5.30)$$

From (5.23, 5.25, 5.26, 5.30) and Lemma 5.2

$$\mathcal{D}_E^\varepsilon(\lambda) = \sum_{n=1}^{\infty} \mathcal{D}_E^\varepsilon(\lambda_n) = \sum_{n=1}^{\infty} [\mathcal{C}_{\varepsilon n T}(\lambda_n^+, \lambda_n^-) + \varepsilon n E T |\lambda_n|] = \sum_{n=1}^{\infty} \sum_{j=0}^{n-1} (\mathcal{C}_{\varepsilon T}(\lambda_n^j, \lambda_n^{j+1}) + \varepsilon E T |\lambda_n|) \quad . \quad (5.31)$$

Let now

$$\mu^{\varepsilon, E} = \varepsilon \sum_{n=1}^{\infty} \sum_{j=1}^{n-1} \lambda_n^j \quad .$$

Note that

$$\mu^{\varepsilon, E} = \varepsilon \sum_{n=1}^{\infty} \sum_{j=0}^n \lambda_n^j - \varepsilon \sum_{n=1}^{\infty} \lambda_n^0 - \varepsilon \sum_{n=1}^{\infty} \lambda_n^n \quad .$$

By (5.26, 5.29, 5.27) we obtain

$$|\mu^{\varepsilon, E}| = \varepsilon \sum_{n=1}^{\infty} (n+1) |\lambda_n^\pm| - 2\varepsilon |\lambda^\pm| = 1 \implies \mu^{\varepsilon, E} \in \mathcal{M}_1^+ \quad . \quad (5.32)$$

By (5.26, 5.29)

$$\begin{aligned} \sum_{n=1}^{\infty} \sum_{j=0}^{n-1} \mathcal{C}_{\varepsilon T}(\lambda_n^j, \lambda_n^{j+1}) &\geq \mathcal{C}_{\varepsilon T} \left(\sum_{n=1}^{\infty} \sum_{j=0}^{n-1} \lambda_n^j, \sum_{n=1}^{\infty} \sum_{j=1}^n \lambda_n^{j+1} \right) = \varepsilon^{-1} \mathcal{C}_{\varepsilon T} \left(\varepsilon \sum_{n=1}^{\infty} \sum_{j=0}^{n-1} \lambda_n^j, \varepsilon \sum_{n=1}^{\infty} \sum_{j=1}^n \lambda_n^{j+1} \right) \\ &= \varepsilon^{-1} \mathcal{C}_{\varepsilon T} (\mu^{\varepsilon, E} + \varepsilon \lambda^+, \mu^{\varepsilon, E} + \varepsilon \lambda^-) \quad . \quad (5.33) \end{aligned}$$

From (5.27, 5.31, 5.33 , 5.32)

$$\mathcal{D}_E^\varepsilon(\lambda) - \langle T \rangle^\varepsilon E \geq \varepsilon^{-1} \mathcal{C}_{\varepsilon T}(\mu^{\varepsilon, E} + \varepsilon\lambda^+, \mu^{\varepsilon, E} + \varepsilon\lambda^-) \geq \varepsilon^{-1} \inf_{\mu \in \mathcal{M}_1^+} \mathcal{C}_{\varepsilon T}(\mu + \varepsilon\lambda^+, \mu + \varepsilon\lambda^-) . \quad (5.34)$$

Finally, Theorem 1-3, (5.24, 5.28, 5.34) imply

$$\widehat{\mathcal{C}}_T(\lambda) \geq \mathcal{D}_E(\lambda) - TE = \lim_{\varepsilon \rightarrow 0} \mathcal{D}_E^\varepsilon(\lambda) - \langle T \rangle^\varepsilon E \geq \limsup_{\varepsilon \rightarrow 0} \varepsilon^{-1} \inf_{\mu \in \mathcal{M}_1^+} \mathcal{C}_{\varepsilon T}(\mu + \varepsilon\lambda^+, \mu + \varepsilon\lambda^-) .$$

□

5.2 Proof of theorem 2

From Theorem 1- (1) we get

$$\widehat{\mathcal{C}}_{\varepsilon T}(\varepsilon\lambda) = \varepsilon \widehat{\mathcal{C}}_T(\lambda) .$$

We now apply Lemma 5.4, adapted to the case where $|\lambda^\pm| := \int \lambda^\pm \neq 1$. Then

$$\mathcal{C}_T(\lambda^+, \lambda^-) = |\lambda^\pm| \mathcal{C}_T\left(\frac{\lambda^+}{|\lambda^+|}, \frac{\lambda^-}{|\lambda^-|}\right) \geq |\lambda^\pm| \widehat{\mathcal{C}}_T\left(\frac{\lambda}{|\lambda^\pm|}\right) = \widehat{\mathcal{C}}_{T/|\lambda^\pm|}(\lambda) .$$

Note that $\int d\mu + \varepsilon d\lambda^\pm = 1 + O(\varepsilon)$, hence

$$\varepsilon^{-1} \mathcal{C}_{\varepsilon T}(\mu + \varepsilon\lambda^+, \mu + \varepsilon\lambda^-) \geq \widehat{\mathcal{C}}_{T_\varepsilon}(\lambda)$$

where $T_\varepsilon \rightarrow T$ as $\varepsilon \rightarrow 0$. Hence

$$\liminf_{\varepsilon \rightarrow 0} \inf_{\mathcal{M}_1^+} \varepsilon^{-1} \mathcal{C}_{\varepsilon T}(\mu + \varepsilon\lambda^+, \mu + \varepsilon\lambda^-) \geq \widehat{\mathcal{C}}_T(\lambda) .$$

The Theorem follows from this and Lemma 5.8.

□

5.3 Proof of Theorem 4

We have to show that for any $(\mu, \lambda) \in \mathcal{M}_1^+ \times \mathcal{M}_0$ and any sequence $(\mu_n, \lambda_n) \rightarrow (\mu, \lambda)$ as $n \rightarrow \infty$:

$$\liminf_{n \rightarrow \infty} n \mathcal{C}_{T/n}(\mu_n + n^{-1}\lambda_n^+, \mu_n + n^{-1}\lambda_n^-) \geq \widehat{\mathcal{C}}(\lambda \parallel \mu) \quad (5.35)$$

and, in addition, *there exists* a sequence $(\hat{\mu}_n, \hat{\lambda}_n) \rightarrow (\mu, \lambda)$ for which

$$\lim_{n \rightarrow \infty} n \mathcal{C}_{T/n}(\hat{\mu}_n + n^{-1}\hat{\lambda}_n^+, \hat{\mu}_n + n^{-1}\hat{\lambda}_n^-) = \widehat{\mathcal{C}}(\lambda \parallel \mu) . \quad (5.36)$$

The inequality (5.35) follows directly from Lemma 5.6. To prove (5.36), we first consider the sequence $(\tilde{\mu}_n, \tilde{\lambda}_n)$ subjected to Lemma 5.5. From Lemma 5.7 and Lemma 5.5,

$$\lim_{j \rightarrow \infty} \limsup_{n \rightarrow \infty} n \mathcal{C}_{T/n}(\tilde{\mu}_j + n^{-1}\tilde{\lambda}_j^+, \tilde{\mu}_j + n^{-1}\tilde{\lambda}_j^-) \leq \lim_{j \rightarrow \infty} \widehat{\mathcal{C}}_T(\tilde{\lambda}_j \parallel \tilde{\mu}_j) = \widehat{\mathcal{C}}(\lambda \parallel \mu) .$$

So, there exists a subsequence j_n along which

$$\limsup_{n \rightarrow \infty} n\mathcal{C}_{T/n} \left(\tilde{\mu}_{j_n} + n^{-1}\tilde{\lambda}_{j_n}^+, \tilde{\mu}_{j_n} + n^{-1}\tilde{\lambda}_{j_n}^- \right) \leq \widehat{\mathcal{C}}(\lambda \|\mu) .$$

This, with (5.35), implies (5.36).

The second part of the theorem follows from (5.35) and Theorem 2.

□

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