

A TRANSPORT INEQUALITY ON THE SPHERE OBTAINED BY MASS TRANSPORT

DARIO CORDERO-ERAUSQUIN

Using McCann's transportation map, we establish a transport inequality on compact manifolds with positive Ricci curvature. This inequality contains the sharp spectral comparison estimates.

1. Introduction

Extending the mass transportation approach to sharp Sobolev-type inequalities from Euclidean space to curved geometries remains a challenging problem. In the present note, we propose a new twist in the classical transportation technique that allows for a transport inequality which contains sharp Poincaré inequalities.

The method applies to a (compact) Riemannian manifold of dimension $n \geq 2$ having a lower bound on the Ricci curvature of the form $\text{Ric} \geq (n-1)k^2g$ with $k > 0$ and g the Riemannian metric. By scaling the distances, we can always assume that $k = 1$.

So, in the rest of the paper $M = (M, g)$ will stand for an n -dimensional Riemannian manifold satisfying

$$(1) \quad \text{Ric} \geq (n-1)g.$$

The main example is the usual sphere $S^n \subset \mathbb{R}^{n+1}$. The interest, perhaps, in stating a result under the condition (1), even if one aims at the sphere only, is that it makes it clear that we will not use any of the algebraic properties of the sphere. Our computations are modeled on the sphere case; the extension to the situation given by (1) relies on Bishop comparison's estimates only. We will denote by $d\sigma = d\text{vol}/\text{vol}(M)$ the Riemannian volume measure normalized to be a probability measure. The distance will be denoted d ; recall as well that M has diameter smaller than π .

A simple but important result is that, on such manifold M , the spectral gap for the Laplacian satisfies $\lambda_1 \geq n$. Equivalently, one has the following Wirtinger–Poincaré

MSC2010: primary 49Q20, 53C21; secondary 60E15.

Keywords: transport inequality, Ricci curvature, optimal mass transport.

inequality: for every Lipschitz function g on M ,

$$(2) \quad \text{Var}_\sigma(g) := \int \left(g - \int g \, d\sigma \right)^2 d\sigma \leq \frac{1}{n} \int |\nabla g|^2 \, d\sigma.$$

The L^2 proof of this inequality as done by Lichnerowicz using Bochner's formula is rather short and elementary. In the particular case of the sphere, one can also use the expansion of g in the spherical harmonics basis; moreover, in this case, equality holds for linear functions, which are eigenfunctions for the spherical Laplacian.

It is well known that Poincaré inequalities are not well suited to mass transport techniques. However, in the Euclidean case and under appropriate curvature assumptions, one can prove very easily using mass transport (Brenier map) *stronger* inequalities such as *transport inequalities* or *logarithmic Sobolev inequalities* (see [Cordero-Erausquin 2002]). So it is quite annoying that no mass transport proof of the sharp log-Sobolev inequality (see [Ledoux 2000]), say, is available on M . Indeed, the straightforward adaptation of the techniques from Euclidean space leads to a log-Sobolev inequality with a constant $(n - 1)$ in place of the expected constant n . Similarly, the transport inequality (definitions are recalled below) that one gets by standard techniques is as follows: for every $f \geq 0$ on M with $\int f \, d\sigma = 1$,

$$(3) \quad \mathcal{W}_c(f \, d\sigma, \sigma) \leq \int f \log f \, d\sigma$$

for the cost $c(d) := \frac{1}{2}(n - 1)d^2$. Linearization of this inequality gives only a weak form of (2) with $1/(n - 1)$ in place of the correct $1/n$. Let us note that by an abstract result of Otto and Villani [2000], the log-Sobolev inequality mentioned above with the sharp constant n implies that the transport inequality (3) holds with the cost $c(d) := \frac{1}{2}nd^2$. As for the log-Sobolev inequality, it is not known how to reach this inequality using mass transport.

The difficulty is to properly quantify the interplay between dimension and nonzero curvature in the mass transportation techniques.

This was partly overcome in [Lott and Villani 2007] (see also [Villani 2009, Chapter 20 and 21]). There, the authors manage to prove some Sobolev-like inequalities under the so called “curvature-dimension condition $CD(K, n)$ ” that imply, after linearization, sharp spectral bounds. To be precise, their assumption is that the metric measure space (M, d, σ) satisfies a curvature-dimension lower bound which is defined in terms of uniform convexity along optimal transport of a class of entropy functionals. From this assumption, they deduce a (not very natural) Sobolev-like inequality. This inequality has no reason to be sharp when the curvature is nonzero, but after linearization it gives the correct Poincaré inequality (2) (so in a sense it is sharp at first order). Of course, it is known, by the properties of optimal transport on manifolds (McCann's map), that a Riemannian manifold with

condition (1) satisfies the curvature-dimension criterion. So putting all together, we see that Lott and Villani's work is already an answer to the question on how to use mass transport to derive some sharp dimensional inequalities. But of course, it is rather indirect, and no standard inequality that one could prove using optimal transport on a manifold is easy to extract from it. Actually, this is somehow the content of Open Problem 21.11 in [Villani 2009].

Our original motivation was to provide, in the particular case of a manifold, a different, more direct, approach based on the geometric properties of McCann's transport map. The aim was to find an inequality that contained the sharp bound (2). Eventually, we managed to establish a new, suitable, transport inequality, that is an inequality between an entropy functional and a transportation cost functional (we recommend the survey [Gozlan and Léonard 2010] for background on transport inequalities). The question of obtaining the sharp log-Sobolev inequality using mass transport remains.

Let us introduce the following classical dimensional entropy: given a probability density f on M , meaning a Borel nonnegative function on M with $\int f d\sigma = 1$, we put

$$H_{n,\sigma}(f) := n \int (f - f^{1-1/n}) d\sigma = n - n \int f^{1-1/n} d\sigma.$$

Note that $H_{n,\sigma}$ is a nonnegative convex functional of f .

We will consider transportation costs given by functions of the distance d on M . Given a function $c : \mathbb{R} \rightarrow \mathbb{R}^+$ (or rather $c : [0, \pi] \rightarrow \mathbb{R}^+$ in our case), the associated Kantorovich transportation cost between two Borel probability measures μ and ν on M is defined by

$$\mathcal{W}_c(\mu, \nu) := \inf_{\pi} \iint c(d(x, y)) d\pi(x, y)$$

where the infimum is taken over all probability measures π on $M \times M$ projecting on μ and ν , respectively.

In the proof of the Theorem below, will use McCann's map, which arises from an optimizer in the functional \mathcal{W}_c when c is the quadratic cost, $c(d) = d^2/2$; we shall recall McCann's result in detail later. However, let us emphasize that, although we will use this quadratic-optimal map, the cost in our transport inequality will be a different function of the distance.

Our cost function is defined for $d \in [0, \pi)$ by

$$c_n(d) := n - \frac{\sin^{n-1} d}{S_n(d)^{n-1}} - (n-1) \frac{S_n(d)}{\tan d}$$

and at the limit by $c_n(\pi) = +\infty$, where S_n is the familiar function defined for

$d \in [0, \pi]$ by

$$S_n(d) := \left(n \int_0^d \sin^{n-1} s \, ds \right)^{1/n}.$$

We have, as expected, $c_n(0) = 0$ (since $S_n(t) \sim t$ at 0) and $c_n(d) > 0$ for $d > 0$.

We now state the transport inequality satisfied by the uniform measure σ on M .

Theorem. *Let M be an n -dimensional Riemannian manifold with positive Ricci curvature satisfying (1) and let σ be its normalized Riemannian volume. Then, for every probability density f on M we have*

$$\mathcal{W}_{c_n}(f \, d\sigma, \sigma) \leq H_{n,\sigma}(f).$$

We will see that the cost $c_n(d(x, y))$ behaves like $(n - 1)d(x, y)^2/2$ for small distances, so it may seem that we are back to the bad situation (3) where we were stuck with the constant $(n - 1)$. However, the entropy $H_{n,\sigma}$ is *better*, i.e., smaller, than the usual entropy $\int f \log f \, d\sigma$ (note that $H_{n,\sigma}(f) \nearrow \int f \log(f) \, d\sigma$ as $n \rightarrow +\infty$), and as a matter of fact we will reproduce the sharp Poincaré inequality. So there is an interesting trade-off between the cost and the entropy. Incidentally, both sides of our inequality are zero when $n = 1$ (which is a good sign), meaning that we don't derive any result on the torus S^1 , although it might be possible, by looking at first orders when $n \rightarrow 1$ and analyzing the proof below, to guess what one should get in this case.

The next section contains the proof of the Theorem. In the last section we give some properties of the cost c_n and we explain how to derive the sharp spectral gap inequality (2) from the Theorem.

2. Proof of the theorem

We start by recalling the result of [McCann 2001]. Given two (compactly supported) probability densities f and g on a manifold M with respect to $d\text{vol}$, the Riemannian volume, there exists a Lipschitz function $\theta : M \rightarrow \mathbb{R}$ such that $-\theta$ is *c-concave* and the map

$$T(x) = \exp_x(\nabla\theta(x))$$

pushes forward $f \, d\text{vol}$ to $g \, d\text{vol}$. The latter means that for every (bounded or nonnegative) Borel function u on M ,

$$\int u(y)g(y) \, d\text{vol}(y) = \int u(T(x))f(x) \, d\text{vol}(x).$$

The *c-concavity* of $-\theta$ is defined by the property that there exists a Lipschitz function ψ such that $-\theta(x) = \inf_y \{\psi(y) + d(x, y)^2/2\}$. This implies (and is formally equivalent to) that at every point x where θ is differentiable, and thus

$y := T(x)$ is uniquely defined, the function $v \rightarrow \theta(v) + \frac{1}{2}d(v, y)^2 - \frac{1}{2}d(x, y)^2$ achieves its minimum at $v = x$.

Following a classical approach, the map T is constructed by establishing that $\pi = (\text{Id} \times T)_* f d\text{vol}$ is the optimizer for $\mathcal{W}_c(f d\text{vol}, g d\text{vol})$ when c is the quadratic cost. We will not use this property, though.

As explained in [Cordero-Erausquin et al. 2001, 2006], it is possible to do, in a weak sense, the change of variable $y = T(x)$ and to establish a pointwise Jacobian change of variable equation. To be precise, let us set, whenever it makes sense,

$$dT_x := Y(H + \text{Hess}_x \theta)$$

where, for fixed $x \in M$, the linear operators $Y : T_x M \rightarrow T_{T(x)} M$ and $H : T_x M \rightarrow T_x M$ are defined by

$$Y := d(\exp_x)_{\nabla \theta(x)} \quad \text{and} \quad H := \text{Hess}_x d_{T(x)}^2 / 2,$$

with the notation $d_y(\cdot) = d(y, \cdot)$ for fixed $y \in M$. Then, one has

$$f(x) = g(T(x)) \det dT_x \quad (f d\text{vol})\text{-a.e.}$$

The set of points where this equation holds is contained in the set of $x \in M$ where θ is differentiable at x with $\gamma(t) := \exp_x(t \nabla \theta(x))$ being the unique minimizing geodesic between $x = \gamma(0)$ and $T(x) = \gamma(1) \notin \text{cut}(x)$, and such that $\text{Hess}_x \theta$ exists, in the sense of Aleksandrov for the Lipschitz (and locally semiconvex) function θ ; later we shall use that $\text{tr Hess } \theta =: \Delta \theta \leq \Delta_{\mathcal{D}} \theta$, where $\Delta_{\mathcal{D}} \theta$ is the distributional Laplacian of the Lipschitz function θ . The c -concavity of $-\theta$ then implies the following, crucial monotonicity property of T , which holds $(f d\text{vol})\text{-a.e.}$:

$$(4) \quad H + \text{Hess } \theta \geq 0.$$

In Euclidean space, $H = \text{Id}$ and we recover that $T(x) = x + \nabla \theta$ is the gradient of the convex function $|x|^2/2 + \theta(x)$ — the Brenier map.

We refer the interested (or worried) reader to [Cordero-Erausquin et al. 2001, 2006] where these facts are carefully stated and proved.

So, under the assumptions of the theorem, let $T(x) = \exp_x(\nabla \theta)$ be the McCann map pushing σ forward to $f d\sigma$. Denote the displacement distance by

$$\alpha(x) := d(x, T(x)) = |\nabla \theta(x)| \in [0, \pi].$$

The Jacobian equation satisfied almost everywhere is then

$$(5) \quad f(T(x))^{-1} = \det(Y(H + \text{Hess}_x \theta))$$

with $Y := d(\exp_x)_{\nabla \theta(x)}$ and $H := \text{Hess}_x d_{T(x)}^2 / 2$.

For $x \in M$ a point where Equation (5) holds, let $E_1 := \nabla\theta/|\nabla\theta|$ be the direction of transport, completed by E_2, \dots, E_n in order to have an orthonormal frame. In this basis, the symmetric operator H takes the form

$$\begin{pmatrix} 1 & 0 \\ 0 & K \end{pmatrix}$$

and the classical Bishop comparison estimates (see [Petersen 1998], for example) ensure that under (1) we have

$$\det Y \leq \left(\frac{\sin \alpha}{\alpha} \right)^{n-1} =: v_n(\alpha)^n \quad \text{and} \quad \operatorname{tr} K \leq (n-1) \frac{\alpha}{\tan \alpha} =: w_n(\alpha).$$

Of course, these inequalities are equalities when $M = S^n$, a case where Y and K can be computed explicitly (see [Cordero-Erausquin 1999]).

If we write $\operatorname{Hess}_x \theta = \begin{pmatrix} a & b^t \\ b & M \end{pmatrix}$, where M is a symmetric $(n-1) \times (n-1)$ matrix and $a := \operatorname{Hess}_x \theta(E_1) \cdot E_1$ (all the quantities depend on x , of course), then we have

$$\begin{aligned} f(T(x))^{-1} &= \det \left[Y \begin{pmatrix} 1+a & b^t \\ b & K+M \end{pmatrix} \right] \leq v_n(\alpha)^n \det \begin{pmatrix} 1+a & b^t \\ b & K+M \end{pmatrix} \\ &\leq v_n(\alpha)^n \det \begin{pmatrix} 1+a & 0 \\ 0 & K+M \end{pmatrix} \\ &= v_n(\alpha)^n \det \begin{pmatrix} (1+a)\mu(\alpha)^{-(n-1)} & 0 \\ 0 & \mu(\alpha)K + \mu(\alpha)M \end{pmatrix}, \end{aligned}$$

where μ is a numerical C^1 positive function defined on $[0, \pi]$ that will be fixed later. Note that $1+a \geq 0$ and $K+M \geq 0$ by (4). Using the arithmetic-geometric inequality, namely $\det^{1/n} \leq \operatorname{tr}/n$ on nonnegative matrices, we then get that

$$nf(T(x))^{-1/n} \leq v_n(\alpha) \left((1+a)\mu(\alpha)^{-(n-1)} + \mu(\alpha)w_n(\alpha) + \mu(\alpha)(\Delta\theta - a) \right).$$

We integrate this inequality with respect to σ . Integration by parts gives

$$\int v_n(\alpha)\mu(\alpha)\Delta\theta d\sigma \leq - \int (v_n\mu)'(\alpha)\nabla\alpha \cdot \nabla\theta d\sigma.$$

When θ is smooth, the previous equation is an equality, but as we explained above, the Laplacian we used is smaller than the distributional Laplacian in general.

By construction, $\nabla\alpha \cdot \nabla\theta = \alpha \operatorname{Hess}\theta(E_1) \cdot E_1 = \alpha a$ (that this property should be used to improve mass transportation techniques on manifolds was suggested to

us by Michael Schmuckenschläger: personal communication, 2001). So we find

$$\begin{aligned} n \int f^{1-1/n} d\sigma &\leq \int (v_n(\alpha)\mu(\alpha)^{-(n-1)} - \mu(\alpha)v_n(\alpha) - \alpha \cdot (v_n\mu)'(\alpha))a d\sigma \\ &\quad + \int (\mu(\alpha)^{-(n-1)} + \mu(\alpha)w_n(\alpha))v_n(\alpha) d\sigma. \end{aligned}$$

We now want to choose the numerical function μ such that for all $t \in [0, \pi]$,

$$(6) \quad v_n(t)\mu(t)^{-(n-1)} - \mu(t)v_n(t) - t(v_n\mu)'(t) = 0.$$

Setting $h(t) := t\mu(t)v_n(t)$, the previous equation rewrites as

$$h'(t) = v_n(t)(h(t)/tv_n(t))^{-(n-1)} = v_n(t)^n t^{n-1} h(t)^{-(n-1)},$$

or equivalently

$$\frac{1}{n}(h^n)'(t) = \sin^{n-1} t,$$

which suggests the choice $h = S_n$. So the function *defined* by $\mu(t) := S_n(t)/tv_n(t)$ satisfies (6), and consequently we have the desired inequality:

$$n \int f^{1-1/n} d\sigma \leq \int \left(\frac{\sin^{n-1} \alpha(x)}{S_n(\alpha(x))^{n-1}} + (n-1) \frac{S_n(\alpha(x))}{\tan \alpha(x)} \right) d\sigma(x). \quad \square$$

3. Further remarks

We start with some properties of the function

$$c_n(\alpha) = n - \frac{\sin^{n-1} \alpha}{S_n(\alpha)^{n-1}} - (n-1) \frac{S_n(\alpha)}{\tan \alpha}, \quad \alpha \in [0, \pi].$$

First, observe that for $\alpha \in [0, \pi]$,

$$\int_0^\alpha \sin^{n-1} s \cos s ds \leq \int_0^\alpha \sin^{n-1} s ds \leq \int_0^\alpha s^{n-1} ds$$

so that

$$\sin \alpha \leq S_n(\alpha) \leq \alpha.$$

This implies that $c_n \geq 0$. It also gives that $0 \leq (\alpha - S_n(\alpha))/\alpha^2 \leq (\alpha - \sin \alpha)/\alpha^2$ and consequently, for $\alpha \rightarrow 0$,

$$S_n(\alpha) = \alpha + o(\alpha^2).$$

In turn, this gives the behavior of $c_n(\alpha)$ when $\alpha \rightarrow 0$:

$$(7) \quad c_n(\alpha) \sim (n-1)\alpha^2/2.$$

To perform this series expansion of c_n , write $S_n(\alpha) = \alpha + a\alpha^3 + o(\alpha^3)$; the coefficient a indeed disappears in the second order. We believe (from numerical examples) that

the function c_n is convex on $[0, \pi]$. But since we don't need this property (which seems a bit more technical), we leave this question for another time.

It is well known that the property (7) of the cost is sufficient to derive by linearization, from the corresponding transport inequality, a Poincaré-type inequality. The standard procedure is to first state an infimal convolution inequality (for the Hamilton–Jacobi semigroup), obtained by dualizing the transportation cost and the entropy, and then to linearize (see [Gozlan and Léonard 2010]). Actually, it is enough to dualize only the transportation cost (we don't want to dualize the entropy, since eventually we will linearize it).

Recall the classical Kantorovich duality: for two probability measures μ and ν on M and for a cost c ,

$$\mathcal{W}_c(\mu, \nu) = \sup_{\varphi} \left\{ \int Q_c(\varphi) d\mu - \int \varphi d\nu \right\}$$

where the supremum is taken over all (Lipschitz) functions $\varphi : M \rightarrow \mathbb{R}$ and

$$Q_c(\varphi)(x) := \inf_{y \in M} \{ \varphi(y) + c(d(x, y)) \} \quad \text{for all } x \in M.$$

Note that $Q_c(\varphi) \leq \varphi$ (provided $c \geq 0$ and $c(0) = 0$) and that the bigger the cost is in terms of $d(x, y)$, the closer $Q_c(\varphi)$ is to φ .

Let g be a smooth function on M with $\int g d\sigma = 0$, and $\varepsilon > 0$ small. Applying our transport inequality to the probability density $f = 1 + \varepsilon \lambda g$ where $\lambda > 0$ is a constant to be fixed later, and using the above-mentioned duality with the test function $\varphi = \varepsilon g$ we get

$$(8) \quad \int Q_{c_n}(\varepsilon g)(1 + \varepsilon \lambda g) d\sigma - \int (\varepsilon g) d\sigma \leq H_{n,\sigma}(1 + \varepsilon \lambda g).$$

On one hand we have, for the entropy term, uniformly on M ,

$$n((1 + \varepsilon \lambda g) - (1 + \varepsilon \lambda g)^{1-1/n}) = \varepsilon \lambda g + \varepsilon^2 \frac{n-1}{2n} (\lambda g)^2 + o(\varepsilon^2).$$

On the other hand, because of (7) we have

$$Q_{c_n}(\varepsilon g) = \varepsilon \left(g - \varepsilon \frac{1}{2(n-1)} |\nabla g|^2 + o(\varepsilon) \right).$$

Putting these two expansions in (8), we see that the orders 0 and 1 vanish (they have to, since the constant function $\mathbf{1}$ is an equality case in the transport inequality), and the inequality between the second orders reads as

$$\left(\lambda - \frac{n-1}{2n} \lambda^2 \right) \int g^2 d\sigma \leq \frac{1}{2(n-1)} \int |\nabla g|^2 d\sigma.$$

Picking $\lambda = \frac{n}{n-1}$ we get the sharp Poincaré inequality $\int g^2 d\sigma \leq \frac{1}{n} \int |\nabla g|^2 d\sigma$.

References

- [Cordero-Erausquin 1999] D. Cordero-Erausquin, “Inégalité de Prékopa–Leindler sur la sphère”, *C. R. Acad. Sci. Paris Sér. I Math.* **329**:9 (1999), 789–792. MR 2000k:26022 Zbl 0945.26026
- [Cordero-Erausquin 2002] D. Cordero-Erausquin, “Some applications of mass transport to Gaussian-type inequalities”, *Arch. Ration. Mech. Anal.* **161**:3 (2002), 257–269. MR 2003h:49076 Zbl 0998.60080
- [Cordero-Erausquin et al. 2001] D. Cordero-Erausquin, R. J. McCann, and M. Schmuckenschläger, “A Riemannian interpolation inequality à la Borell, Brascamp and Lieb”, *Invent. Math.* **146**:2 (2001), 219–257. MR 2002k:58038 Zbl 1026.58018
- [Cordero-Erausquin et al. 2006] D. Cordero-Erausquin, R. J. McCann, and M. Schmuckenschläger, “Prékopa–Leindler type inequalities on Riemannian manifolds, Jacobi fields, and optimal transport”, *Ann. Fac. Sci. Toulouse Math.* (6) **15**:4 (2006), 613–635. MR 2008j:49111 Zbl 1125.58007
- [Gozlan and Léonard 2010] N. Gozlan and C. Léonard, “Transport inequalities: a survey”, *Markov Process. Related Fields* **16**:4 (2010), 635–736. MR 2895086 Zbl 1229.26029 arXiv 1003.3852
- [Ledoux 2000] M. Ledoux, “The geometry of Markov diffusion generators”, *Ann. Fac. Sci. Toulouse Math.* (6) **9**:2 (2000), 305–366. MR 2002a:58045 Zbl 0980.60097
- [Lott and Villani 2007] J. Lott and C. Villani, “Weak curvature conditions and functional inequalities”, *J. Funct. Anal.* **245**:1 (2007), 311–333. MR 2008f:53039 Zbl 1119.53028
- [McCann 2001] R. J. McCann, “Polar factorization of maps on Riemannian manifolds”, *Geom. Funct. Anal.* **11**:3 (2001), 589–608. MR 2002g:58017 Zbl 1011.58009
- [Otto and Villani 2000] F. Otto and C. Villani, “Generalization of an inequality by Talagrand and links with the logarithmic Sobolev inequality”, *J. Funct. Anal.* **173**:2 (2000), 361–400. MR 2001k:58076 Zbl 0985.58019
- [Petersen 1998] P. Petersen, *Riemannian geometry*, Graduate Texts in Mathematics **171**, Springer, New York, 1998. 2nd ed. published in 2006. MR 98m:53001 Zbl 0914.53001
- [Villani 2009] C. Villani, *Optimal transport: old and new*, Grundlehren der Mathematischen Wissenschaften [Fundamental Principles of Mathematical Sciences] **338**, Springer, Berlin, 2009. MR 2010f:49001 Zbl 1156.53003

Received February 6, 2013. Revised August 3, 2013.

DARIO CORDERO-ERAUSQUIN
 INSTITUT DE MATHÉMATIQUES DE JUSSIEU
 UNIVERSITÉ PIERRE ET MARIE CURIE (PARIS 6)
 4 PLACE JUSSIEU
 75252 PARIS CEDEX 5
 FRANCE
 dario.cordero@imj-prg.fr