

## COMPARISON BETWEEN $W_2$ DISTANCE AND $\dot{H}^{-1}$ NORM, AND LOCALIZATION OF WASSERSTEIN DISTANCE \*

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**Abstract.** It is well known that the quadratic Wasserstein distance  $W_2(\cdot, \cdot)$  is formally equivalent, for infinitesimally small perturbations, to some weighted  $H^{-1}$  homogeneous Sobolev norm. In this article I show that this equivalence can be integrated to get non-asymptotic comparison results between these distances. Then I give an application of these results to prove that the  $W_2$  distance exhibits some localization phenomenon: if  $\mu$  and  $\nu$  are measures on  $\mathbb{R}^n$  and  $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}_+$  is some bump function with compact support, then under mild hypotheses, you can bound above the Wasserstein distance between  $\varphi \cdot \mu$  and  $\varphi \cdot \nu$  by an explicit multiple of  $W_2(\mu, \nu)$ .

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### 1. FOREWORD

This article is divided into two sections, each of which having its own introduction. Section 2 deals with general results of comparison between Wasserstein distance and homogeneous Sobolev norm, while Section 3 handles an application to localization of  $W_2$  distance.

### 2. NON-ASYMPTOTIC EQUIVALENCE BETWEEN $W_2$ DISTANCE AND $\dot{H}^{-1}$ NORM

#### 2.1. Introduction

In all this section,  $M$  denotes a connected Riemannian manifold endowed with its distance  $dist(\cdot, \cdot)$  and its standard measure  $\lambda$  provided by the volume form (so, in the case  $M = \mathbb{R}^n$ ,  $\lambda$  is the Lebesgue measure). Let us give a few standard definitions which will be at the core of our work:

- For  $\mu, \nu$  two positive measures on  $M$ , denoting by  $\Pi(\mu, \nu)$  the set of (positive) measures on  $M \times M$  whose respective marginals are  $\mu$  and  $\nu$ , for  $\pi \in \Pi(\mu, \nu)$  one defines

$$I(\pi) := \int_{M \times M} dist(x, y)^2 \pi(dx, dy) \quad (2.1)$$

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and then

$$W_2(\mu, \nu) := \inf\{I(\pi) \mid \pi \in \Pi(\mu, \nu)\}^{1/2}. \tag{2.2}$$

$W_2$  is a (possibly infinite) distance, called the *quadratic Wasserstein distance* ([13], Sect. 7.1). Note that this distance is finite only between measures having the same total mass.

- On the other hand, for  $\mu$  a (positive) measure on  $M$ , if  $f$  is a  $C^1$  real function on  $M$ , one denotes

$$\|f\|_{\dot{H}^1(\mu)} := \left( \int_M |\nabla f(x)|^2 \mu(dx) \right)^{1/2}, \tag{2.3}$$

which defines a semi-norm; for  $\nu$  a signed measure on  $M$ , one then denotes

$$\|\nu\|_{\dot{H}^{-1}(\mu)} := \sup\{|\langle f, \nu \rangle| \mid \|f\|_{\dot{H}^1(\mu)} \leq 1\}, \tag{2.4}$$

where the duality product  $\langle f, \nu \rangle$  denotes the integral of the function  $f$  against the measure  $\nu$ .<sup>2</sup> We observe that  $\|\cdot\|_{\dot{H}^{-1}(\mu)}$  defines a (possibly infinite) norm, which we will call the  $\dot{H}^{-1}(\mu)$  *weighted homogeneous Sobolev norm*. Note that this norm is finite only for measures having zero total mass. In the case  $\mu$  is the standard measure, we will merely write “ $\dot{H}^{-1}$ ” for “ $\dot{H}^{-1}(\lambda)$ ”.

The  $W_2$  Wasserstein distance is an important object in analysis; but it is non-linear, which makes it harder to study. For infinitesimal perturbations however, the linearized behaviour of  $W_2$  is well known: if  $\mu$  is a positive measure on  $M$  and  $d\mu$  is an infinitesimally small perturbation of this measure,<sup>3</sup> one has formally (see [13], Sect 7.6, or [9], Sect. 7)

$$W_2(\mu, \mu + d\mu) = \|d\mu\|_{\dot{H}^{-1}(\mu)} + o(d\mu). \tag{2.5}$$

More precisely, one has the following equality, known as the *Benamou–Brenier formula* ([2], Prop. 1.1) (see [10] when  $M$  is a general Riemannian manifold): for two positive measures  $\mu, \nu$  on  $M$ ,

$$W_2(\mu, \nu) = \inf \left\{ \int_0^1 \|d\mu_t\|_{\dot{H}^{-1}(\mu_t)} \mid \mu_0 = \mu, \mu_1 = \nu \right\}. \tag{2.6}$$

Then, a natural question is the following: are there *non-asymptotic* comparisons between the  $W_2$  distance and the  $\dot{H}^{-1}$  norm? Concretely, we are looking for inequalities like

$$C_a \|\mu - \nu\|_{\dot{H}^{-1}(\mu)} \leq W_2(\mu, \nu) \leq C_b \|\mu - \nu\|_{\dot{H}^{-1}(\mu)} \tag{2.7}$$

for constants  $0 < C_a \leq C_b < \infty$ , under mild assumptions on  $\mu$  and  $\nu$ .

## 2.2. Controlling $W_2$ by $\dot{H}^{-1}$

**Theorem 2.1.** *For any positive measures  $\mu, \nu$  on  $M$ ,*

$$W_2(\mu, \nu) \leq 2 \|\mu - \nu\|_{\dot{H}^{-1}(\mu)}. \tag{2.8}$$

*Proof.* We suppose that  $\|\mu - \nu\|_{\dot{H}^{-1}(\mu)} < \infty$ , otherwise there is nothing to prove. For  $t \in [0, 1]$ , let

$$\mu_t := (1 - t)\mu + t\nu, \tag{2.9}$$

<sup>2</sup>The rationale behind the use of duality notation in this article is that we cannot use the notation “ $d\mu$ ” to refer to the measure of a small volume: see indeed Footnote 3 below.

<sup>3</sup>Beware that here  $d\mu$  denotes a small measure on  $M$ , not the value of  $\mu$  on a small volume.

<sup>4</sup>This formula has to be understood in the sense that, for every measure  $\nu$ , one has  $W_2(\mu, \mu + \varepsilon\nu) \stackrel{\varepsilon \rightarrow 0}{\equiv} |\varepsilon| \|\nu\|_{\dot{H}^{-1}(\mu)} + o(\varepsilon)$ . As explained in the references cited, some regularity assumptions on  $\nu$  shall be required for that property to hold rigorously: in particular, one must have  $\nu \ll \mu$  with a bounded and smooth enough density.

so that  $\mu_0 = \mu$ ,  $\mu_1 = \nu$  and  $d\mu_t = (\mu - \nu)dt$ . Then, by the Benamou–Brenier formula (2.6):

$$W_2(\mu, \nu) \leq \int_0^1 \|\mu - \nu\|_{\dot{H}^{-1}(\mu_t)} dt. \quad (2.10)$$

Now, we use the following key lemma, whose proof is postponed:

**Lemma 2.2.** *If  $\mu, \mu'$  are two measures such that  $\mu' \geq \rho\mu$  for some  $\rho > 0$ , then  $\|\cdot\|_{\dot{H}^{-1}(\mu')} \leq \rho^{-1/2} \|\cdot\|_{\dot{H}^{-1}(\mu)}$ .*<sup>5</sup>

Here obviously  $\mu_t \geq (1-t)\mu$ , so

$$W_2(\mu, \nu) \leq \int_0^1 (1-t)^{-1/2} \|\mu - \nu\|_{\dot{H}^{-1}(\mu)} dt = 2 \|\mu - \nu\|_{\dot{H}^{-1}(\mu)}. \quad (2.11)$$

□

**Corollary 2.3.** *If  $\mu \geq \rho\lambda$  for some  $\rho > 0$ , then*

$$W_2(\mu, \nu) \leq 2\rho^{-1/2} \|\mu - \nu\|_{\dot{H}^{-1}}. \quad (2.12)$$

*Proof.* Just use that  $\|\cdot\|_{\dot{H}^{-1}(\mu)} \leq \rho^{-1/2} \|\cdot\|_{\dot{H}^{-1}}$  by Lemma 2.2. □

*Proof of Lemma 2.2.* Take  $\mu' \geq \rho\mu$  and let  $\nu$  be a signed measure on  $M$  such that  $\mu + \nu$  is positive; then  $\mu' + \rho\nu$  is also positive. For  $m$  a measure on  $M$ , we denote by  $\text{diag}(m)$  the measure on  $M \times M$  supported by the diagonal whose marginals (which are equal) are  $m$ , *i.e.*:

$$(\text{diag}(m))(A \times B) := m(A \cap B); \quad (2.13)$$

with that notation,

$$\pi \in \Pi(\mu, \mu + \nu) \Rightarrow \rho\pi + \text{diag}(\mu' - \rho\mu) \in \Pi(\mu', \mu' + \rho\nu), \quad (2.14)$$

and

$$I(\rho\pi + \text{diag}(\mu' - \rho\mu)) = \rho I(\pi). \quad (2.15)$$

Therefore, taking infima,

$$\begin{aligned} W_2(\mu', \mu' + \rho\nu)^2 &= \inf \{I(\pi') \mid \pi' \in \Gamma(\mu', \mu' + \rho\nu)\} \\ &\leq \inf \{I(\rho\pi + \text{diag}(\mu' - \rho\mu)) \mid \pi \in \Gamma(\mu, \mu + \nu)\} \\ &= \rho \inf \{I(\pi) \mid \pi \in \Gamma(\mu, \mu + \nu)\} = \rho W_2(\mu, \mu + \nu)^2. \end{aligned} \quad (2.16)$$

For infinitesimally small  $\nu$ ,<sup>6</sup> it follows by equation (2.5) that  $\|\rho\nu\|_{\dot{H}^{-1}(\mu')}^2 \leq \rho \|\nu\|_{\dot{H}^{-1}(\mu)}^2$ , hence  $\|\nu\|_{\dot{H}^{-1}(\mu')} \leq \rho^{-1/2} \|\nu\|_{\dot{H}^{-1}(\mu)}$ . This relation remains true even for non-infinitesimal  $\nu$  by linearity, which ends the proof. □

**Remark 2.4.** Lemma 2.2 could also be proved very quickly by using the definition (2.3)–(2.4) of the  $\dot{H}^{-1}(\mu)$  norm. The proof above, however, has the advantage that it does not need the precise expression of  $\|\cdot\|_{\dot{H}^{-1}(\mu)}$ , but only the fact that it is the linearized  $W_2$  distance.

<sup>5</sup>Beware that here ‘ $\cdot$ ’ stands for a *measure*, not for a function: otherwise the formula would be false.— When  $f$  is a function,  $\|f\|_{\dot{H}^{-1}(\mu)}$  stands for the  $\dot{H}^{-1}(\mu)$  norm of the measure having density  $f$  w.r.t.  $\mu$ .

<sup>6</sup>To make rigorous the formal argument of taking an infinitesimally small  $\nu$ , according to Footnote 4 above, one would have to replace  $\nu$  by  $\varepsilon\nu_1$ , where  $\nu_1$  is a regular enough measure, and to let  $\varepsilon$  tend to 0; then the regularity assumption on  $\nu_1$  would be relaxed by a classical approximation argument. Anyway, Lemma 2.2 can also be proved easily and rigorously without referring to optimal transportation at all, *cf.* Remark 2.4 below.

### 2.3. Controlling $\dot{H}^{-1}$ by $W_2$

**Theorem 2.5.** *Assume  $M$  has nonnegative Ricci curvature. Then for any positive measures  $\mu, \nu$  on  $M$  such that  $\mu \leq \rho_0 \lambda$  and  $\nu \leq \rho_1 \lambda$ ,*

$$\|\mu - \nu\|_{\dot{H}^{-1}} \leq \frac{2(\rho_0^{1/2} - \rho_1^{1/2})}{\ln(\rho_0 / \rho_1)} W_2(\mu, \nu). \tag{2.17}$$

(For  $\rho_1 = \rho_0$ , the right-hand side of (2.17) is to be taken as  $\rho_0^{1/2} W_2(\mu, \nu)$  by continuity).

**Remark 2.6.** For  $M = \mathbb{R}^n$  a similar result was already stated in ([7], Prop. 2.8), with a different proof.

*Proof.* Assume that  $W_2(\mu, \nu) < \infty$ , otherwise there is nothing to prove. Let  $(\mu_t)_{0 \leq t \leq 1}$  be the displacement interpolation between  $\mu$  and  $\nu$  (cf. [14], Chapt. 7), which is such that  $\mu_0 = \mu$ ,  $\mu_1 = \nu$  and the infimum in (2.6) is attained with  $\|d\mu_t\|_{\dot{H}^{-1}(\mu_t)} = W_2(\mu, \nu) dt \forall t$ . Since Ricci curvature is nonnegative, the Lott–Sturm–Villani theory tells us that, denoting by  $\|\mu\|_\infty$  the essential supremum of the density of  $\mu$  w.r.t.  $\lambda$ , one has  $\|\mu_t\|_\infty \leq \|\mu_0\|_\infty^{1-t} \|\mu_1\|_\infty^t \leq \rho_0^{1-t} \rho_1^t$  (see [14], Cor. 17.19 or [5], Lem. 6.1); so that  $\|\cdot\|_{\dot{H}^{-1}} \leq \rho_0^{(1-t)/2} \rho_1^{t/2} \|\cdot\|_{\dot{H}^{-1}(\mu_t)}$  by Lemma 2.2.

Then, by the integral triangle inequality for normed vector spaces,

$$\begin{aligned} \|\mu - \nu\|_{\dot{H}^{-1}} &= \left\| \int_0^1 d\mu_t \right\|_{\dot{H}^{-1}} \leq \int_0^1 \|d\mu_t\|_{\dot{H}^{-1}} \leq \int_0^1 \rho_0^{(1-t)/2} \rho_1^{t/2} \|d\mu_t\|_{\dot{H}^{-1}(\mu_t)} \\ &= \left( \int_0^1 \rho_0^{(1-t)/2} \rho_1^{t/2} dt \right) W_2(\mu, \nu) = \frac{2(\rho_0^{1/2} - \rho_1^{1/2})}{\ln(\rho_0 / \rho_1)} W_2(\mu, \nu). \end{aligned} \tag{2.18}$$

□

**Remark 2.7.** Taking into account the dimension  $n$  of the manifold  $M$ , the bound on  $\|\mu_t\|_\infty$  could be refined into

$$\|\mu_t\|_\infty \leq ((1-t)\rho_0^{-1/n} + t\rho_1^{-1/n})^{-n} \tag{2.19}$$

(cf. [8], Thm. 2.3), which would yield a slightly sharper bound in equation (2.17), namely:

$$\|\mu - \nu\|_{\dot{H}^{-1}} \leq \left( \int_0^1 ((1-t)\rho_0^{-1/n} + t\rho_1^{-1/n})^{-n/2} dt \right) W_2(\mu, \nu) = \begin{cases} \frac{\rho_0^{1/2-1/n} - \rho_1^{1/2-1/n}}{(n/2-1)(\rho_1^{-1/n} - \rho_0^{-1/n})} W_2(\mu, \nu) & n \geq 2; \\ \frac{\log(\rho_1 / \rho_0)}{2(\rho_0^{-1/2} - \rho_1^{-1/2})} W_2(\mu, \nu) & n = 2. \end{cases} \tag{2.20}$$

For  $n = 1$  it turns out that one can let tend  $\rho_1 \rightarrow \infty$  in (2.20) without making the integral diverge; which leads to a much more powerful result:

**Theorem 2.8.** *When  $M$  is an interval of  $\mathbb{R}$ , then under the sole assumption that  $\mu \leq \rho_0 \lambda$ , one has for all positive measures  $\nu$  on  $M$ :*

$$\|\mu - \nu\|_{\dot{H}^{-1}} \leq 2\rho_0^{1/2} W_2(\mu, \nu). \tag{2.21}$$

**Remark 2.9.** For  $n \geq 2$  there is no hope to get a bound valid for all  $\nu$ , because then it can occur that  $W_2(\mu, \nu) < \infty$  but  $\|\mu - \nu\|_{\dot{H}^{-1}} = \infty$ : for instance, take  $\mu$  to be the uniform measure on the 2-dimensional sphere and  $\nu$  a Dirac mass.

### 3. APPLICATION TO LOCALIZATION OF WASSERSTEIN DISTANCE

#### 3.1. Introduction

In all this section, we work in the Euclidian space  $\mathbb{R}^n$ , whose norm is denoted by  $|\cdot|$ .  $dist(x, A) := \inf\{|x - y| \mid y \in A\}$  denotes the distance between a point  $x$  and a set  $A$ ;  $A^c$  denotes the complement of  $A$ ;  $\lambda$  denotes the Lebesgue measure. We will use the following notation to handle measures:

- For  $\mu$  a measure on  $\mathbb{R}^n$  and  $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$  a measurable map,  $f_* \mu$  denotes the pushforward of  $\mu$  by  $f$ , that is,  $(f_* \mu)(A) := \mu(f^{-1}(A))$ .
- For  $\mu$  a measure on  $\mathbb{R}^n$  and  $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}_+$  a nonnegative measurable function,  $\varphi \cdot \mu$  denotes the measure such that  $(\varphi \cdot \mu)(dx) := \varphi(x)\mu(dx)$ .

We will also use the following norms on measures:

- $\|\mu\|_{\dot{H}^{-1}(\nu)}$  has the same definition as in Section 2;
- $\|\mu\|_1 := \int_{\mathbb{R}^n} |\mu(dx)|$  is the total variation norm of  $\mu$ ;<sup>7</sup>
- For  $\nu$  a positive measure with  $\mu \ll \nu$ , we define

$$\|\mu\|_{L^2(\nu)} := \left( \int_{\text{supp } \nu} \left( \frac{\mu(dx)}{\nu(dx)} \right)^2 \nu(dx) \right)^{1/2}. \tag{3.1}$$

For  $A \subset \mathbb{R}^n$ , we also denote  $\|\cdot\|_{L^2(A)}$  for  $\|\cdot\|_{L^2(\mathbf{1}_A \cdot \lambda)}$ .

The goal of this section is to give an application of Theorem 2.1 to the problem of *localization* of the quadratic Wasserstein distance. Morally, the question is the following: take two measures  $\mu, \nu$  on  $\mathbb{R}^n$  being close to each other in the sense of  $W_2$  distance; is it true that  $\mu$  and  $\nu$  remain close when you consider their restrictions to a subset of  $\mathbb{R}^n$ ? Concretely, if  $\varphi$  is a non-negative real function on  $\mathbb{R}^n$  with compact support (plus some technical assumptions to be specified later), we want to bound above  $W_2(a\varphi \cdot \mu, \varphi \cdot \nu)$  by some multiple of  $W_2(\mu, \nu)$  — where, in the former expression,  $a$  is a constant factor ensuring that  $a\varphi \cdot \mu$  and  $\varphi \cdot \nu$  have the same mass (for otherwise the distance between  $\varphi \cdot \mu$  and  $\varphi \cdot \nu$  is generically infinite).

This question, which was my initial motivation for the results of Section 2, was asked to me by Xavier TOLSA, who needed such a result for his paper [12] on characterizing uniform rectifiability in terms of mass transport. Actually Xavier managed to devise a proof of his own ([12], Thm. 1.1), but it was quite long (about thirty pages) and involved arguments of multi-scale analysis. With Theorem 2.1 at hand, however, the reasoning becomes far more direct; moreover we will be able to relax some of the assumptions of Xavier’s theorem.

#### 3.2. Statement of the theorem

**Theorem 3.1.** *Let  $\mu, \nu$  be (positive) measures on  $\mathbb{R}^n$  having the same total mass; let  $B$  be a ball of  $\mathbb{R}^n$  (whose radius will be denoted by  $R$  when needed). Assume that on  $B$ , the density of  $\mu$  w.r.t. the Lebesgue measure is bounded above and below:*

$$\exists 0 < m_1 \leq m_2 < \infty \quad \forall x \in B \quad m_1 \lambda(dx) \leq \mu(dx) \leq m_2 \lambda(dx). \tag{3.2}$$

Let  $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}_+$  be a function such that:

- (i)  $\varphi$  is zero outside  $B$ ;
- (ii) There exist  $0 < c_1 \leq c_2 < \infty$  such that for all  $x \in B$ ,  $c_1 dist(x, B^c)^2 \leq \varphi(x) \leq c_2 dist(x, B^c)^2$ .
- (iii)  $\varphi$  is  $k$ -Lipschitz for some  $k < \infty$ .

<sup>7</sup>Note that in the case  $\mu$  is a positive measure on  $\mathbb{R}^n$ , then  $\|\mu\|_1$  is nothing but  $\mu(\mathbb{R}^n)$ .

<sup>8</sup>What we denote here by  $\mu(dx)/\nu(dx)$  here is what is commonly called  $(d\mu/d\nu)(x)$ : indeed, as we already told, in this article we reserve the use of “ $d\mu$ ” to denote a mass distribution of infinitesimally small magnitude, rather than for the mass of an infinitely small volume.

Then, denoting  $a := \|\varphi \cdot \nu\|_1 / \|\varphi \cdot \mu\|_1$ ,

$$W_2(a\varphi \cdot \mu, \varphi \cdot \nu) \leq C(n) \frac{c_2^{3/2} m_2^{3/2}}{c_1^{3/2} m_1^{3/2}} k c_1^{-1/2} W_2(\mu, \nu), \tag{3.3}$$

for  $C(n) < \infty$  some absolute constant only depending on  $n$ . Moreover, one can bound explicitly  $C(n)$  in such a way that  $C(n) = O(n^{1/2})$  when  $n \rightarrow \infty$ .<sup>9</sup>

**Remark 3.2.** Theorem 3.1 relaxes the assumptions of Theorem 1.1 of [12] on the following points: first, Tolsa’s theorem required that  $|\nabla\varphi|$  was bounded by a multiple of  $\text{dist}(\cdot, B^c)$ , while ours does not impose any specific control on  $|\nabla\varphi|$  near the boundary of  $B$ ; second, Tolsa’s theorem worked only for radially symmetric  $\varphi$ . Also, contrary to [12], our conclusions state explicitly how the bound on  $W_2(a\varphi \cdot \mu, \varphi \cdot \nu)$  depends on the constants  $k, c_1, c_2, m_1, m_2$  and on the dimension  $n$ .

**Remark 3.3.** Actually the constraint that the support of  $\varphi$  is a ball is of little importance: we could assume as well that it would be a cube, a simplex, or many other shapes, as the corollary below shows:

**Corollary 3.4.** *Make the same assumptions as in Theorem 3.1, except that  $B$  need not be a ball: instead, we only assume that, denoting by  $B_\circ$  the (true) ball having the same volume as  $B$ , there exists a bijection  $\Phi: B \leftrightarrow B_\circ$  mapping the uniform measure on  $B$  onto the uniform measure on  $B_\circ$  (i.e. such that  $\Phi_*(\mathbf{1}_B \cdot \lambda) = \mathbf{1}_{B_\circ} \cdot \lambda$ ) such that  $\Phi$  is bi-Lipschitz (i.e. such that both  $\Phi$  and  $\Phi^{-1}$  are Lipschitz). Denote by  $\|\Phi\|_{\text{Lip}}$  and  $\|\Phi^{-1}\|_{\text{Lip}}$  the optimal Lipschitz constants for resp.  $\Phi$  and  $\Phi^{-1}$ . Then, the conclusion of Theorem 3.1 remains true, except that now you have to replace the factor  $C(n)$  by*

$$(\|\Phi\|_{\text{Lip}} \|\Phi^{-1}\|_{\text{Lip}})^5 C(n). \tag{3.4}$$

*Proof.* Consider the measures  $\mu_\circ := \Phi_* \mu$  and  $\nu_\circ := \Phi_* \nu$ , and the bump function  $\varphi_\circ := \varphi \circ \Phi^{-1}$ ; then,  $\mu_\circ, \nu_\circ$  and  $\varphi_\circ$  satisfy the original assumptions of Theorem 3.1, the roles of ‘ $m_1$ ’ and ‘ $m_2$ ’ (in the ball situation) being held by  $m_1$  and  $m_2$  (in the general situation) themselves, the role of ‘ $k$ ’ being held by  $\|\Phi^{-1}\|_{\text{Lip}} k$ , and the roles of ‘ $c_1$ ’ and ‘ $c_2$ ’ being held by  $c_1 / \|\Phi\|_{\text{Lip}}^2$  and  $c_2 \|\Phi^{-1}\|_{\text{Lip}}^2$ . Therefore, applying (3.3):

$$W_2(a\varphi_\circ \cdot \mu_\circ, \varphi_\circ \cdot \nu_\circ) \leq C(n) \|\Phi\|_{\text{Lip}}^4 \|\Phi^{-1}\|_{\text{Lip}}^4 \frac{c_2^{3/2} m_2^{3/2}}{c_1^{3/2} m_1^{3/2}} W_2(\mu_\circ, \nu_\circ). \tag{3.5}$$

But the optimal transportation plan from  $\mu$  to  $\nu$ , with cost  $W_2(\mu, \nu)^2$ , can be pushed forward by  $\Phi$  into a (not optimal in general) transportation plan from  $\mu_\circ$  to  $\nu_\circ$ , whose cost will then be  $\leq \|\Phi\|_{\text{Lip}}^2 W_2(\mu, \nu)^2$ ; so  $W_2(\mu_\circ, \nu_\circ) \leq \|\Phi\|_{\text{Lip}} W_2(\mu, \nu)$ . Similarly  $W_2(a\varphi \cdot \mu, \varphi \cdot \nu) \leq \|\Phi^{-1}\|_{\text{Lip}} W_2(a\varphi_\circ \cdot \mu_\circ, \varphi_\circ \cdot \nu_\circ)$ . The announced result follows.  $\square$

### 3.3. Proof of the main theorem

In the sequel we will shorthand  $W_2(\mu, \nu) =: w$ , and also  $\varphi \cdot \mu =: \hat{\mu}$ , resp.  $\varphi \cdot \nu =: \hat{\nu}$ . Let  $g =: \text{id} + S$  be a map achieving optimal transportation from  $\nu$  to  $\mu$ , i.e. such that  $\mu = g_* \nu$  with  $\int_{\mathbb{R}^n} |S(y)|^2 \nu(dy) = w^2$ .<sup>10</sup>

Our strategy will consist in transforming  $\hat{\nu}$  into  $a\hat{\mu}$  according to the following procedure:

- ① We apply the transportation plan  $g$  to  $\hat{\nu}$ ; this transforms  $\hat{\nu}$  into some measure  $\hat{\mu}^*$ . The measure  $\hat{\mu}^*$  is not supported by  $B$  a priori, so we split it into  $\hat{\mu}_B^* + \hat{\mu}_c^* := \mathbf{1}_B \cdot \hat{\mu}^* + \mathbf{1}_{B^c} \cdot \hat{\mu}^*$ .

<sup>9</sup>For instance, with the estimates of this article, one finds that  $C(n) := 47n^{1/2}$  fits—though this may be strongly suboptimal.

<sup>10</sup>Actually such a  $g$  does not always exist, as it can occur that the optimal transportation plan from  $\nu$  to  $\mu$  “splits points” if  $\nu$  is not regular enough. However it would suffice to use the general formalism of transportation plans to handle that case: we do not do it here to keep notation light, but this is straightforward. Also note that it is not obvious that the infimum in (2.2) is attained: again, that is not a real problem as our proof still works by considering a sequence of transportation plans approaching optimality.

- ② Denoting  $a_c := \|\hat{\mu}_c^*\|_1 / \|\hat{\mu}\|_1$ , we then transform  $\hat{\mu}_c^*$  into  $a_c \hat{\mu}$  according to an arbitrary transference plan.
- ③ Finally, denoting  $a_B := \|\hat{\mu}_B^*\|_1 / \|\hat{\mu}\|_1$ ,<sup>11</sup> we transform  $\hat{\mu}_B^*$  into  $a_B \hat{\mu}$  according to the optimal transference plan: the cost of this operation is  $W_2(\hat{\mu}_B^*, a_B \hat{\mu})$ , which we bound above by  $2 \|\hat{\mu}_B^* - a_B \hat{\mu}\|_{\dot{H}^{-1}(a_B \hat{\mu})}$  thanks to Theorem 2.1.

Then, denoting by  $W_2(\textcircled{1}), W_2(\textcircled{2}), W_2(\textcircled{3})$  the respective Wasserstein distances of these steps, we shall have  $W_2(\hat{\nu}, a\hat{\mu}) \leq W_2(\textcircled{1}) + (W_2(\textcircled{2})^2 + W_2(\textcircled{3})^2)^{1/2}$ .

Let us begin with bounding the cost of Step ①. The squared cost of this step is

$$W_2(\textcircled{1})^2 = \int |S(y)|^2 \hat{\nu}(dy) = \int |S(y)|^2 \varphi(y) \nu(dy) \leq \sup \varphi \times \int |S(y)|^2 \nu(dy) = \sup \varphi \times w^2 \leq c_2 R^2 w^2, \tag{3.6}$$

whence  $W_2(\textcircled{1}) \leq c_2^{1/2} R w$ .

Now consider Step ②. As  $a_c \hat{\mu}$  is supported by  $B$ , one has obviously

$$W_2(\textcircled{2})^2 \leq \int_{B^c} (\text{dist}(x, B) + 2R)^2 \hat{\mu}_c^*(dx) = \int_{B^c} (\text{dist}(x, B) + 2R)^2 \hat{\mu}^*(dx). \tag{3.7}$$

From that we deduce that  $W_2(\textcircled{2}) \leq 2c_2^{1/2} R w$  by the following computation:

$$\begin{aligned} \int_{B^c} (\text{dist}(x, B) + 2R)^2 \hat{\mu}^*(dx) &= \int_{g(y) \notin B} (\text{dist}(g(y), B) + 2R)^2 \varphi(y) \nu(dy) \\ &\leq c_2 \int_{\substack{y \in B \\ g(y) \notin B}} (\text{dist}(g(y), B) + 2R)^2 \text{dist}(y, B^c)^2 \nu(dy) \\ &\leq c_2 \int_{\substack{y \in B \\ g(y) \notin B}} (R \text{dist}(g(y), B) + 2R \text{dist}(y, B^c))^2 \nu(dy) \\ &\leq 4c_2 R^2 \int_{\substack{y \in B \\ g(y) \notin B}} (\text{dist}(g(y), B) + \text{dist}(y, B^c))^2 \nu(dy) \\ &\leq 4c_2 R^2 \int |y - g(y)|^2 \nu(dy) = 4c_2 R^2 w^2. \end{aligned} \tag{3.8}$$

Step ③ is the difficult one. We begin with observing that it is easy to bound the  $L^2(B)$  distance between  $\hat{\mu}_B^*$  and  $\hat{\mu}$ : indeed, denoting by  $f =: \text{id} + T$  the inverse map of  $g$ ,<sup>12</sup>

$$\begin{aligned} \|\hat{\mu}_B^* - \hat{\mu}\|_{L^2(\mathbf{1}_B \cdot \mu)}^2 &= \int_B \left( \frac{\hat{\mu}^*(dx) - \varphi(x)\mu(dx)}{\mu(dx)} \right)^2 \mu(dx) = \int_B (\varphi(f(x)) - \varphi(x))^2 \mu(dx) \\ &\leq k^2 \int_{\mathbb{R}^n} |x - f(x)|^2 \mu(dx) = k^2 \int |T(x)|^2 \mu(dx) = k^2 w^2, \end{aligned} \tag{3.9}$$

(where we used that  $\hat{\mu}^*(dx) = \hat{\nu}(d(f(x))) = \varphi(f(x))\nu(d(f(x))) = \varphi(f(x))\mu(dx)$ ), so that

$$\|\hat{\mu}_B^* - \hat{\mu}\|_{L^2(B)}^2 \leq k^2 m_2 w^2. \tag{3.10}$$

<sup>11</sup>Observe that  $a_B + a_c = a$ .

<sup>12</sup>For  $f$  to exist,  $g$  should be bijective, which is not always true *stricto sensu*; but we can safely carry out the reasoning with pretending so, by the same argument as in Footnote 10 on page 1494.

<sup>13</sup>Remember that when  $\nu$  stands for a *measure*,  $\|\nu\|_{L^2(\mu)}$  means what is more commonly denoted by  $\|\text{d}\nu/\text{d}\mu\|_{L^2(\mu)}$ , so that the relation  $\mu \leq m\lambda$  implies that  $\|\nu\|_{L^2(\lambda)}^2 \leq m \|\nu\|_{L^2(\mu)}^2$ —while on the other hand, when  $f$  stands for a *function*, one has  $\|f\|_{L^2(\mu)}^2 \leq m \|f\|_{L^2(\lambda)}^2$ .

Now we have to link  $\|\cdot\|_{L^2(B)}$  with  $\|\cdot\|_{\dot{H}^{-1}(\mu)}$ . This is achieved by the following lemma, whose proof is postponed:

**Lemma 3.5.** *Define  $\hat{\lambda}$  to be the measure on  $B$  such that  $\hat{\lambda}(dx) := \text{dist}(x, B^c)^2 \lambda(dx)$ . Then, for any signed measure  $m$  on  $B$  having total mass zero:*

$$\|m\|_{\dot{H}^{-1}(\hat{\lambda})} \leq C_1(n)^{1/2} \|m\|_{L^2(B)}, \tag{3.11}$$

where  $C_1(n)$  is some absolute constant only depending on  $n$ . Moreover, taking  $C_1(n) := ((2e + 1)n - 1) \vee 8e$  fits.

Thanks to Theorem 2.1 and Lemma 3.5, we have that

$$W_2(\textcircled{3}) \leq 2 \|a_B \hat{\mu} - \hat{\mu}_B^*\|_{\dot{H}^{-1}(a_B \hat{\mu})} \leq 2(a_B c_1 m_1)^{-1/2} \|a_B \hat{\mu} - \hat{\mu}_B^*\|_{\dot{H}^{-1}(\hat{\lambda})} \leq 2C_1(n)^{1/2} (a_B c_1 m_1)^{-1/2} \|a_B \hat{\mu} - \hat{\mu}_B^*\|_{L^2(B)}. \tag{3.12}$$

Next, we compute

$$\begin{aligned} \|a_B \hat{\mu} - \hat{\mu}_B^*\|_{L^2(B)} &= \left\| \frac{\|\hat{\mu}_B^*\|_1}{\|\hat{\mu}\|_1} \hat{\mu} - \hat{\mu}_B^* \right\|_{L^2(B)} \leq \frac{|\|\hat{\mu}_B^*\|_1 - \|\hat{\mu}\|_1|}{\|\hat{\mu}\|_1} \|\hat{\mu}\|_{L^2(B)} + \|\hat{\mu}_B^* - \hat{\mu}\|_{L^2(B)} \\ &\leq \frac{\|\hat{\mu}\|_{L^2(B)}}{\|\hat{\mu}\|_1} \|\hat{\mu}_B^* - \hat{\mu}\|_1 + \|\hat{\mu}_B^* - \hat{\mu}\|_{L^2(B)} \leq \left( \frac{\|\hat{\mu}\|_{L^2(B)}}{\|\hat{\mu}\|_1} \lambda(B)^{1/2} + 1 \right) \|\hat{\mu}_B^* - \hat{\mu}\|_{L^2(B)} \\ &\leq \left( \frac{c_2 m_2}{c_1 m_1} \frac{\lambda(B)^{1/2} \|\hat{\lambda}\|_{L^2(B)}}{\|\hat{\lambda}\|_1} + 1 \right) \|\hat{\mu}_B^* - \hat{\mu}\|_{L^2(B)} \stackrel{14}{\leq} (\sqrt{6} \frac{c_2 m_2}{c_1 m_1} + 1) \|\hat{\mu}_B^* - \hat{\mu}\|_{L^2(B)} \\ &\stackrel{(3.10)}{\leq} (\sqrt{6} + 1) \frac{c_2 m_2}{c_1 m_1} k m_2^{1/2} w, \end{aligned} \tag{3.13}$$

so that, combining (3.12) and (3.13), we have got:

$$W_2(\textcircled{3}) \leq (2\sqrt{6} + 2) C_1(n)^{1/2} a_B^{-1/2} \frac{c_2 m_2^{3/2}}{c_1 m_1^{3/2}} \frac{k}{c_1^{1/2}} w. \tag{3.14}$$

Equation (3.14) is the kind of bound we were looking for, *provided*  $a_B \lesssim 1$ . Though this will be the case in practice (since we are mainly interested in cases where  $\nu$  is close to  $\mu$  and thus  $\hat{\mu}^*$  is close to  $\hat{\mu}$ ), this is not quite satisfactory yet. So, what can we do when  $a_B \ll 1$ , that is, when  $\|\hat{\mu}_B^*\|_1 \ll \|\hat{\mu}\|_1$ ? In fact that case is easier, because transportation between small measures has low cost, while  $w$  has to be large to make  $\hat{\mu}_B^*$  very different from  $\hat{\mu}$ .

The computations are the following. First, it is obvious that

$$W_2(\textcircled{3}) = W_2(\hat{\mu}_B^*, a_B \hat{\mu}) \leq 2R \|\hat{\mu}_B^*\|_1^{1/2}. \tag{3.15}$$

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<sup>14</sup>This step comes from the computation  $\lambda(B)^{1/2} \|\hat{\lambda}\|_{L^2(B)} / \|\hat{\lambda}\|_1 = (\int_0^1 r^{n-1} dr)^{1/2} (\int_0^1 (1-r)^4 r^{n-1} dr)^{1/2} / (\int_0^1 (1-r)^2 r^{n-1} dr) = (6(1+n)(2+n) / (3+n)(4+n))^{1/2} \leq \sqrt{6} \forall n$ .



Next, observing that  $\varphi(f(x)) \geq \frac{c_1}{c_2} \varphi(x) - 2c_1 \operatorname{dist}(x, B^c) |T(x)|$ ,<sup>15</sup> we compute that

$$\begin{aligned} \|\hat{\mu}_B^*\|_1 &= \int_B \varphi(f(x)) \mu(dx) \geq \int_B \left( \frac{c_1}{c_2} \varphi(x) - 2c_1 \operatorname{dist}(x, B^c) |T(x)| \right) \mu(dx) \\ &\geq \frac{c_1}{c_2} \|\hat{\mu}\|_1 - 2c_1 \left( \int_B \operatorname{dist}(x, B^c)^2 \mu(dx) \right)^{1/2} \left( \int_B |T(x)|^2 \mu(dx) \right)^{1/2} \\ &= \frac{c_1}{c_2} \|\hat{\mu}\|_1 - 2c_1 \|\operatorname{dist}(\cdot, B^c)^2 \cdot \mu\|_1^{1/2} w \geq \frac{c_1}{c_2} \|\hat{\mu}\|_1 - 2c_1 m_2^{1/2} \|\hat{\lambda}\|_1^{1/2} w, \end{aligned} \tag{3.17}$$

whence

$$w \geq \frac{\left( \frac{c_1}{c_2} \|\hat{\mu}\|_1 - \|\hat{\mu}_B^*\|_1 \right)^+}{2c_1 m_2^{1/2} \|\hat{\lambda}\|_1^{1/2}} = \frac{\left( \frac{c_1}{c_2} - a_B \right)^+ \|\hat{\mu}\|_1}{2c_1 m_2^{1/2} \|\hat{\lambda}\|_1^{1/2}} \geq \frac{m_1^{1/2}}{2c_1^{1/2} m_2^{1/2}} \left( \frac{c_1}{c_2} - a_B \right)^+ \|\hat{\mu}\|_1^{1/2}. \tag{3.18}$$

So,

$$W_2(\textcircled{3}) \leq 2R \|\hat{\mu}_B^*\|_1^{1/2} = 2R a_B^{1/2} \|\hat{\mu}\|_1^{1/2} \leq 4R c_1^{1/2} \frac{m_2^{1/2}}{m_1^{1/2}} \frac{a_B^{1/2}}{\left( \frac{c_1}{c_2} - a_B \right)^+} w. \tag{3.19}$$

In the end, choosing either (3.14) if  $a_B \geq c_1 / 2c_2$  or (3.19) if  $c_1 / 2c_2$ , and observing that  $c_1 \leq kR^{-1}$ , one has always:

$$W_2(\textcircled{3}) \leq ((4\sqrt{3} + 2\sqrt{2})C_1(n)^{1/2} \vee 4\sqrt{2}) \frac{c_2^{3/2} m_2^{3/2}}{c_1^{3/2} m_1^{3/2}} \frac{k}{c_1^{1/2}} w. \tag{3.20}$$

**Remark 3.6.** To bound  $W_2(\textcircled{3})$  in the situation where  $a_B \ll 1$ , we could also have started from “ $\varphi(f(x)) \geq \varphi(x) - k|T(x)|$ ” (instead of “ $\varphi(f(x)) \geq \frac{c_1}{c_2} \varphi(x) - 2c_1 \operatorname{dist}(x, B^c) |T(x)|$ ”) to get another bound analogous to (3.17). Following such an approach, the factor  $(c_2 / c_1)^{3/2}$  in (3.19) would be improved into  $(c_2 / c_1)$  in the analogous formula; however the dimensional factor would behave in  $O(n)$  rather than in  $O(n^{1/2})$ .

### 3.4. Proof of Lemma 3.5

It still remains to prove Lemma 3.5, whose statement we recall to be:

**Lemma 3.7.** Denoting  $\hat{\lambda} := \operatorname{dist}(\cdot, B^c)^2 \cdot \lambda$ , one has, for any signed measure  $m$  on  $B$  having total mass zero:

$$\|m\|_{\dot{H}^{-1}(\hat{\lambda})} \leq (((2e + 1)n - 1) \vee 8e)^{1/2} \|m\|_{L^2(B)}. \tag{3.21}$$

—In the sequel, “ $((2e + 1)n - 1) \vee 8e$ ” will be shorthanded into “ $C_1(n)$ ”.

**Remark 3.8.** The bound (3.21) is within a constant factor of being optimal, uniformly in  $n$ , as one sees by taking a linear function  $f$  in (3.24).

*Proof of the lemma.* We begin with translating the lemma into a functional analysis statement by a duality argument. Recall the duality definition of  $\|m\|_{\dot{H}^{-1}(\hat{\lambda})}$  from Section 2:

$$\|m\|_{\dot{H}^{-1}(\hat{\lambda})} := \sup\{|\langle f, m \rangle| \mid \|f\|_{\dot{H}^1(\hat{\lambda})} \leq 1\}. \tag{3.22}$$

<sup>15</sup>This follows from the computation:

$$\varphi(f(x)) \geq c_1 \operatorname{dist}(f(x), B^c)^2 \geq c_1 ((\operatorname{dist}(x, B^c) - |T(x)|)^+)^2 \geq c_1 \operatorname{dist}(x, B^c)^2 - 2c_1 \operatorname{dist}(x, B^c) |T(x)| \geq \frac{c_1}{c_2} \varphi(x) - 2c_1 \operatorname{dist}(x, B^c) |T(x)|. \tag{3.16}$$

There is a similar duality formula for  $\|m\|_{L^2(B)}$ :

$$\|m\|_{L^2(B)} = \sup\{|\langle f, m \rangle| \mid \|f\|_{L^2(B)} \leq 1\}, \tag{3.23}$$

where, for  $f$  a function,  $\|f\|_{L^2(B)}$  has its usual meaning, namely  $\|f\|_{L^2(B)} := (\int_B f(x)^2 \lambda(dx))^{1/2}$ . Since  $m$  is assumed to have total mass zero,  $|\langle f, m \rangle|$  does not change when one adds a constant to  $f$ . On the other hand, when  $f$  describes the set  $\{\|f_0 + a\| \mid a \in \mathbb{R}\}$ ,  $\|f\|_{L^2(B)}$  is minimal when  $a$  is such that  $f$  has zero mean on  $B$ , while the value of  $\|f\|_{\dot{H}^1(\lambda)}$  remains constant.<sup>16</sup> As a consequence, we can restrict the supremum in (3.22) and (3.23) to those  $f$  having zero mean on  $B$ . Thus, the lemma will be implied<sup>17</sup> by proving that

$$\langle f, \mathbf{1}_B \cdot \lambda \rangle = 0 \quad \Rightarrow \quad \|f\|_{L^2(B)} \leq C_1(n)^{1/2} \|f\|_{\dot{H}^1(\lambda)}. \tag{3.24}$$

Going back to the definitions of  $\|\cdot\|_{\dot{H}^{-1}(\lambda)}$  and  $\|\cdot\|_{L^2(B)}$ , relaxing the condition on  $f$  to be centred by projecting it orthogonally in  $L^2(B)$  onto the subspace of centred functions, and denoting by  $P$  the uniform probability measure on  $B$ , Equation (3.24) turns into:

$$\forall f \quad \text{Var}_P(f) \leq C_1(n) \int \text{dist}(x, B^c)^2 |\nabla f(x)|^2 P(dx), \tag{3.25}$$

which we recognize to be a so-called ‘‘improved Poincaré inequality’’ [3, 6]. In general, Poincaré inequalities, bounding the variance of  $f$  by a quadratic integral of its first derivative, are linked with the exponential convergence of a certain diffusion Markov process towards equilibrium (cf. [1], Chap. 2): that probabilistic vision initially guided me to tackle Equation (3.25), although this will not be apparent in the sequel.

To prove (3.25), the first key idea (inspired by [4]) is to separate radial and spherical coordinates. This is, considering the bijection

$$\begin{aligned} \varphi: (0, R) \times \mathbb{S}^{n-1} &\rightarrow B \setminus \{0\} \\ (r, \theta) &\mapsto r\theta \end{aligned} \tag{3.26}$$

(the origin of space being set at the center of  $B$ ), we introduce the measure  $\tilde{P} := \varphi^{-1} \ast P$ , which is obviously the product measure  $\tilde{P}_r \otimes \tilde{P}_\theta$ , where  $\tilde{P}_r$  is the probability measure on  $(0, R)$  such that  $\tilde{P}_r(dr) := nR^{-n}r^{n-1}dr$ , resp.  $\tilde{P}_\theta$  is the uniform measure on the sphere  $\mathbb{S}^{n-1}$ . With this notation, we perform a change of variables to see that (3.25) is equivalent to proving that, for all  $g \in L^2(\tilde{P})$ :

$$C_1(n)^{-1} \text{Var}_{\tilde{P}}(g) \leq \int_0^R \int_{\mathbb{S}^{n-1}} (R-r)^2 (|\nabla_r g(r, \theta)|^2 + r^{-2} |\nabla_\theta g(r, \theta)|^2) \tilde{P}_r(dr) \tilde{P}_\theta(d\theta), \tag{3.27}$$

where  $\nabla_r$  and  $\nabla_\theta$  denote the gradient along resp. the  $r$  coordinate and the  $\theta$  coordinate<sup>18</sup>. We will denote the right-hand side of (3.27) by  $\mathcal{E}(g, g)$ .

Because  $\tilde{P} = \tilde{P}_r \otimes \tilde{P}_\theta$ , we know that  $L^2(\tilde{P})$  can be seen as (the closure of) the tensor product of  $L^2(\tilde{P}_r)$  and  $L^2(\tilde{P}_\theta)$ :

$$L^2(\tilde{P}) = \text{cl}(L^2(\tilde{P}_r) \overset{\perp}{\otimes} L^2(\tilde{P}_\theta)), \tag{3.28}$$

where the symbol  $\overset{\perp}{\otimes}$  means that the Hilbertian structure of  $L^2(\tilde{P})$  is compatible with the Hilbertian structures of  $L^2(\tilde{P}_r)$  and  $L^2(\tilde{P}_\theta)$ —i.e., that  $\langle h_a \otimes u_a, h_b \otimes u_b \rangle_{L^2(\tilde{P})} = \langle h_a, h_b \rangle_{L^2(\tilde{P}_r)} \times \langle u_a, u_b \rangle_{L^2(\tilde{P}_\theta)}$ . Now consider the spherical harmonics  $Y_0, Y_1, \dots$ , which by definition are an orthonormal basis, in  $L^2(\tilde{P}_\theta)$ , of eigenfunctions of the

<sup>16</sup>Here we implicitly assume that  $\int_B |f(x)| \lambda(dx) < \infty$ , which is legitimate since an approximation argument allows to restrict the suprema in (3.22) and (3.23) to those  $f$  having a  $C^\infty$  continuation on  $\text{cl}(B)$ .

<sup>17</sup>Actually there is even equivalence.

<sup>18</sup>In the latter case, we have to use the Riemannian definition of the gradient on  $\mathbb{S}^{n-1}$ .

Laplace–Beltrami operator  $\Delta$  on  $\mathbb{S}^{n-1}$ ; and call  $\ell_0, \ell_1, \dots$  the associated eigenvalues, which are known to be such that (up to permuting indices)  $Y_0 \equiv 1$  with  $\ell_0 = 0$ , and  $\ell_i \leq -(n-1) \forall i \neq 0$  (see for instance [11]). By construction,  $L^2(\tilde{P}_\theta) = \text{cl} \left( \bigoplus_{i \in \mathbb{N}} (\mathbb{R} \cdot Y_i) \right)$ ; therefore, one has that

$$L^2(\tilde{P}) = \text{cl} \left( \bigoplus_{i \in \mathbb{N}} L^2(\tilde{P}_r) \cdot Y_i \right); \tag{3.29}$$

in other words, the functions of  $L^2(\tilde{P})$  are those of the form

$$g(r, \theta) = \sum_{i \in \mathbb{N}} h_i(r) Y_i(\theta), \tag{3.30}$$

with  $\sum_i \|h_i\|_{L^2(\tilde{P}_r)}^2 < \infty$ , and the correspondence is bijective. An interesting point is that, then, one has:

$$\text{Var}_{\tilde{P}}(g) = \text{Var}_{\tilde{P}_r}(h_0) + \sum_{i \neq 0} \|h_i\|_{L^2(\tilde{P}_r)}^2. \tag{3.31}$$

On the other hand, one has

$$\mathcal{E}(g, g) = -\langle Lg, g \rangle_{L^2(\tilde{P})}, \tag{3.32}$$

where

$$(Lg)(r, \theta) := (R-r)^2 \Delta_r g + \left( (n-1) \frac{(R-r)^2}{r} - 2(R-r) \right) e_r \cdot \nabla_r g + \frac{(R-r)^2}{r^2} \Delta_\theta g. \tag{3.33}$$

From (3.33) we see that, since the  $Y_i$  are eigenfunctions of  $\Delta_\theta$ , all the  $L^2(\tilde{P}_r) \cdot Y_i$  are invariant by  $L$ , and that one has:

$$\mathcal{E}(g, g) = \sum_{i \in \mathbb{N}} \int_0^R \left( (R-r)^2 |\nabla h_i(r)|^2 - \ell_i \frac{(R-r)^2}{r^2} h_i(r)^2 \right) \tilde{P}_r(dr). \tag{3.34}$$

So, proving (3.27) becomes equivalent to proving that both following formulas hold for all  $h \in L^2(\tilde{P}_r)$ :

$$\text{Var}_{\tilde{P}_r}(h) \leq C_1(n) \int_0^R (R-r)^2 |\nabla h(r)|^2 \tilde{P}_r(dr); \tag{3.35}$$

$$\|h\|_{L^2(\tilde{P}_r)}^2 \leq C_1(n) \int_0^R \left( (R-r)^2 |\nabla h(r)|^2 + (n-1) \frac{(R-r)^2}{r^2} h(r)^2 \right) \tilde{P}_r(dr). \tag{3.36}$$

Let us start with (3.35). In all the sequel of the proof, we introduce

$$b := 1 - n^{-1}. \tag{3.37}$$

By the Cauchy–Schwarz inequality, one has, for all  $r \in (bR, R)$ :

$$\begin{aligned} (h(r) - h(bR))^2 &= \left( \int_{bR}^r h'(s) ds \right)^2 \leq \left( \int_{bR}^r (R-s)^{-3/2} ds \right) \times \int_{bR}^r (R-s)^{3/2} |\nabla h(s)|^2 ds \\ &\leq 2 \left( (R-r)^{-1/2} - (R-bR)^{-1/2} \right) \int_{bR}^r (R-s)^{3/2} |\nabla h(s)|^2 ds \\ &\leq 2(R-r)^{-1/2} \int_{bR}^r (R-s)^{3/2} |\nabla h(s)|^2 ds. \end{aligned} \tag{3.38}$$

Integrating and using Fubini’s formula, it follows that

$$\begin{aligned}
 \int_{bR}^R (h(r) - h(bR))^2 \tilde{P}_r(dr) &\leq 2 \int_{s=bR}^R \left( \int_{r=s}^R nR^{-n} (R-r)^{-1/2} r^{n-1} dr \right) (R-s)^{3/2} |\nabla h(s)|^2 ds \\
 &\leq 2 \int_{s=bR}^R \left( \int_{r=s}^R nR^{-n} (b^{-1}s)^{n-1} (R-r)^{-1/2} dr \right) (R-s)^{3/2} |\nabla h(s)|^2 ds \\
 &= 2b^{-(n-1)} \int_{s=bR}^R \left( \int_{r=s}^R (R-r)^{-1/2} dr \right) (R-s)^{3/2} |\nabla h(s)|^2 \tilde{P}_r(ds) \\
 &= 4b^{-(n-1)} \int_{s=bR}^R (R-s)^2 |\nabla h(s)|^2 ds. \tag{3.39}
 \end{aligned}$$

One can apply the same line of reasoning for  $r \in (0, bR)$ : the (unweighted this time) Cauchy–Schwarz inequality then yields  $(h(r) - h(bR))^2 \leq (bR - r) \int_r^{bR} |\nabla h(s)|^2 ds$ , whence:

$$\begin{aligned}
 \int_0^{bR} (h(r) - h(bR))^2 \tilde{P}_r(dr) &\leq \int_{s=0}^{bR} \left( \int_{r=0}^s nR^{-n} (bR-r)r^{n-1} dr \right) |\nabla h(s)|^2 ds \\
 &\leq bR^{-(n-1)} \int_{s=0}^{bR} \left( \int_{r=0}^s nr^{n-1} dr \right) |\nabla h(s)|^2 ds = bR \int_0^{bR} |\nabla h(s)|^2 s^n ds \\
 &\leq bn^{-1} R^2 \int_0^{bR} |\nabla h(s)|^2 \tilde{P}_r(ds) \leq b(1-b)^{-2} n^{-1} \int_0^{bR} (R-s)^2 |\nabla h(s)|^2 \tilde{P}_r(ds). \tag{3.40}
 \end{aligned}$$

Summing (3.39) and (3.40), we get that

$$\int_0^R (h(r) - h(bR))^2 \tilde{P}_r(dr) \leq (4b^{-(n-1)} \vee b(1-b)^{-2} n^{-1}) \int_0^R (R-s)^2 |\nabla h(s)|^2 \tilde{P}_r(ds), \tag{3.41}$$

where  $(4b^{-(n-1)} \vee b(1-b)^{-2} n^{-1})$  can itself be bounded by  $((n-1) \vee 4e)$ . The left-hand-side of (3.41) being an upper bound for  $\text{Var}_{\tilde{P}_r}(h)$ , this proves (3.35).

Now we turn to (3.36). For  $r \in (bR, R)$  we have, similarly to (3.38), that

$$(h(r) - h(br))^2 \leq 2(R-r)^{-1/2} \int_{br}^r (R-s)^{3/2} |\nabla h(s)|^2 ds, \tag{3.42}$$

so that

$$h(r)^2 \leq 2h(br)^2 + 4(R-r)^{-1/2} \int_{br}^r (R-s)^{3/2} |\nabla h(s)|^2 ds. \tag{3.43}$$

Then, integrating and applying Fubini’s formula:

$$\int_{bR}^R h(r)^2 \tilde{P}_r(dr) \leq 2 \int_{bR}^R h(br)^2 \tilde{P}_r(dr) + 4 \int_{s=b^2R}^R \left( \int_{r=s \vee bR}^{b^{-1}s \wedge R} nR^{-n} r^{n-1} (R-r)^{-1/2} dr \right) (R-s)^{3/2} |\nabla h(s)|^2 ds. \tag{3.44}$$

By change of variables, the first term of the right-hand side of (3.44) is equal to  $2b^{-n} \int_{b^2R}^{bR} h(s)^2 \tilde{P}_r(ds)$ , which we can bound by

$$2b^{-(n-2)} \frac{(1-b)^{-2}}{n-1} \int_{b^2R}^{bR} (n-1) \frac{(R-r)^2}{r^2} h(s)^2 \tilde{P}_r(ds) \leq 2ne \int_0^R (n-1) \frac{(R-r)^2}{r^2} h(s)^2 \tilde{P}_r(ds). \tag{3.45}$$

The second term of the right-hand side of (3.44) is itself bounded by

$$4b^{-(n-1)} \int_{s=b^2R}^R \left( \int_{r=s}^R (R-r)^{-1/2} dr \right) (R-s)^{3/2} |\nabla h(s)|^2 \tilde{P}_r(ds) \leq 8e \int_0^R (R-s)^2 |\nabla h(s)|^2 \tilde{P}_r(ds). \quad (3.46)$$

This way, we have bounded  $\int_{bR}^R h(r)^2 \tilde{P}_r(dr)$ .

On the other hand, it is trivial that, for  $r \leq bR$ ,

$$h(r)^2 \leq \frac{b^2}{(n-1)(1-b)^2} \times (n-1) \frac{(R-r)^2}{r^2} h(r)^2, \quad (3.47)$$

whence:

$$\int_0^{bR} h(r)^2 \tilde{P}_r(dr) \leq (n-1) \int_0^R (n-1) \frac{(R-r)^2}{r^2} h(r)^2 \tilde{P}_r(dr). \quad (3.48)$$

Combining (3.45), (3.46) and (3.48), we finally get the wanted bound (3.36).  $\square$

**Remark 3.9.** At the time I wrote that proof I was not aware of the already existing results on improved Poincaré inequalities, in particular ([6], Thm. 1.3), which equation (3.25) is actually a particular case of; nor were the people whom I had asked about such inequalities. Compared to the result of [6] however, my equation (3.25) states an explicit value for the constant in the inequality, which moreover is within a constant factor of being optimal, uniformly in the dimension  $n$ ; also, it uses a quite different proof, which may be interesting  $\odot$

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