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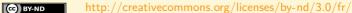
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STABILITY ESTIMATES FOR THE SHARP SPECTRAL GAP BOUND UNDER A CURVATURE-DIMENSION CONDITION

by Max FATHI, Ivan GENTIL & Jordan SERRES (*)

ABSTRACT. — We study stability of the sharp spectral gap bounds for metric-measure spaces satisfying a curvature bound. Our main result, new even in the smooth setting, is a sharp quantitative estimate showing that if the spectral gap of an RCD(N-1,N) space is almost minimal, then the pushforward of the measure by an eigenfunction associated with the spectral gap is close to a Beta distribution. The proof combines estimates on the eigenfunction obtained via a new L^1 -functional inequality for RCD spaces with Stein's method for distribution approximation. We also derive analogous, almost sharp, estimates for infinite and negative values of the dimension parameter.

RÉSUMÉ. — Nous étudions la stabilité du trou spectral pour des espaces métriques mesurés vérifiant un critère de courbure dimension. Notre résultat principal, nouveau même dans le cadre d'un espace lisse, est une estimation quantitative montrant que si le trou spectral d'un espace $\mathrm{RCD}(N-1,N)$ est presque minimal alors l'image par la fonction propre associée au trou spectral de la mesure de référence est proche d'une loi Beta. La preuve combine des estimations de la fonction propre par le biais d'une nouvelle inégalité fonctionnelle de type L^1 dans un espace RCD et la méthode de Stein pour comparer des lois. Nous obtenons aussi des bornes presque optimales pour des espaces ayant une dimension intrinsèque infinie ou bien négative.

1. Introduction

Our goal in this work is to study the stability of sharp spectral gap bounds for Markov diffusion operators \mathcal{L} satisfying a curvature lower bound.

Keywords: Spectral gap, Poincaré inequalities, RCD spaces, Curvature-dimension condition, Stein method.

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Formally (and rigorously in the smooth setting), these operators satisfy both the Bochner inequality, (or curvature-dimension condition CD(K, N)),

(1.1)
$$\frac{1}{2}\mathcal{L}\Gamma(f) - \Gamma(f, \mathcal{L}f) \geqslant K\Gamma(f) + \frac{1}{N}(\mathcal{L}f)^2$$

for all smooth functions, and the diffusion property: for any smooth bounded function ϕ

(1.2)
$$\mathcal{L}\phi(f) = \phi'(f)\mathcal{L}f + \phi''(f)\Gamma(f).$$

This definition has been proposed in the seminal paper [6] by the way of the operator $\Gamma_2(f) := \frac{1}{2}\mathcal{L}\Gamma(f) - \Gamma(f,\mathcal{L}f)$.

An alternative definition of the curvature-dimension condition (now known as the Lott–Sturm–Villani theory [33, 49, 50]) is to require geodesic convexity properties of the entropy in the space of probability measures. In the setting of smooth Riemanian manifolds endowed with a reference probability measure, the curvature-dimension condition is equivalent to a lower bound on the weighted Ricci curvature tensor. However, in the non-smooth setting, requiring the inequality (1.1) to hold pointwise is slightly too strong. We shall be interested in so-called RCD spaces, that satisfy the diffusion property, as well as a weak, integral form of the curvature-dimension condition [3, 18]. This definition is equivalent to the more classical definition via convexity properties of the entropy functional as in the Lott–Sturm–Villani theory when the space is infinitesimally Hilbertian.

For a smooth (C^2 and complete) N-dimensional Riemannian manifold endowed with its Laplacian, if the Ricci curvature tensor is bounded from below by N-1 (which implies the CD(N-1, N) condition), we have a lower bound on the spectral gap, or first positive eigenvalue of the Laplacian:

$$(1.3) \lambda_1(-\Delta) \geqslant N.$$

This bound is sharp, since equality holds for the N-sphere of radius 1. Moreover, this bound is *rigid*: equality holds iff the manifold is isometric to the N-sphere [40]. This phenomenon is related to rigidity results for other sharp bounds, such as on the diameter or the volume.

The next question is whether the bound is stable: if a manifold with Ricci curvature bounded from below by N-1 has a spectral gap close to N, is it close in some sense to the N-sphere? The answer to this question is negative [4]. To get closeness in Gromov–Hausdorff distance, Aubry [5] showed that we must ask that the N-th eigenvalue is close to N, which is true for the sphere, since the multiplicity of N as an eigenvalue is N+1,

and that λ_{N-1} being close to N is not enough. This improved on an earlier result of Petersen [42] who showed that $\lambda_{N+1} \approx N$ suffices. The reason for this phenomenon is that smooth manifolds are not quite the right setting for the problem: it is possible to extend the notion of Ricci curvature lower bounds to non-smooth weighted manifolds, and in that setting there are spaces other than the N-sphere for which equality holds in (1.3), namely spherical suspensions. We refer to [29] for their description.

When it is only λ_1 that is close to N, Cheng [15] showed that the diameter is close to π , and Croke [16] proved the converse statement, still in the smooth, unweighted setting. Later, Bertrand [9] showed that $\lambda_k \approx N$ for $k \leq N$ implies that the manifold contains a piece that is close to \mathbb{S}^k .

We shall now discuss the literature in the non-smooth setting of RCD spaces. This curvature condition, that we shall present in details in Section 2, extends Ricci curvature bounds, and can be introduced using either the Bakry–Ledoux gradient estimates, a weak form of the Bochner inequality or convexity properties of the entropy along Wasserstein geodesics. This setting also makes sense when N is not an integer. The sharp spectral gap estimate for RCD(N-1,N) spaces was proved in [18], and cases of equality were fully described in [28, 29]. More recently, Cavaletti and Mondino tackled rigidity and stability results for geometric comparison theorems in the non-smooth setting using the so-called needle decomposition, a technique pioneered by Klartag [30] to reduce such problems to one-dimensional statements. Most relevant to the present work is the quantitative Obata theorem of [13], which states that on an essentially non-branching CD(N-1,N) space (M,d,μ) with spectral gap λ_1 and associated normalized eigenfunction u there exists a point $x_0 \in M$ such that

$$||u - \sqrt{N+1}\cos(d(\cdot, x_0))||_2 \le C(N)(\lambda_1 - N)^{1/(8N+4)};$$

 $\pi - \operatorname{diam}(M) \le C(N)(\lambda_1 - N)^{1/N}.$

A variant of the diameter estimate was obtained in [27] in the RCD setting, still with a dimension-dependent exponent. We shall include a comparison between the technical estimates on the eigenfunction used in [13] and those used here in Section 3.1. The main differences between Theorem 1.1 below and the results of [13], beyond the norms used being different, is that on the upside we get a dimension-independent exponent in our main estimate, but with the downside of requiring the RCD condition rather than the more general CD condition. Topological sphere theorems were also considered in the RCD setting, in [25], and an averaged version of the maximal diameter theorem was proved in [19], where the only optimizer is the N-sphere.

Our main result is a sharp quantitative estimate on how far the distribution of the pushforward measure by an eigenfunction is of being a beta distribution. More precisely, we show that

THEOREM 1.1. — Let (M, d, μ) be an RCD(N-1, N) space with N > 1, unit mass and spectral gap $\lambda_1 \leq N + \varepsilon$ for some $\varepsilon > 0$. Let f be an eigenfunction of the Laplacian, with eigenvalue λ_1 , and normalized so that $\|\Gamma(f)\|_1 = N/(N+1)$. There is a constant C(N) > 0 (independent of M and f) such that the L^1 -Wasserstein distance between the pushforward of μ by f and a symmetrized Beta distribution with parameters (N/2, N/2) is smaller than $C(N)\varepsilon$.

The order of magnitude ε in the bound is *sharp*, as can be checked by considering an N-sphere of radius $1 - \varepsilon$. The choice of the value of the normalization for $\|\Gamma(f)\|_1$ is to match the value for coordinates on a unit sphere in dimension N, and without the pushforward would simply be close to a scaled Beta distribution.

Symmetrized Beta distributions (with parameters (N/2, N/2)) have densities proportional to $(1-x^2)^{N/2-1}$ on [-1,1]. They appear in this statement because they are precisely the distribution of coordinates on unit spheres, so our statement can be viewed as saying that the space contains a piece that is close to a piece of the N-sphere (when the parameter N is an integer of course).

Our method follows an approach developed in [10] for proving stability of the sharp spectral gap estimate on $RCD(1,\infty)$ spaces, where the model space is the Gauss space. It combines quantitative estimates on eigenfunctions with Stein's method for comparing probability distributions via approximate integration by parts formulas. We shall actually sharpen the quantitative bound of order $\varepsilon^{1/2}$ of [10, Theorem 1.3] in the $RCD(1,\infty)$ setting to $\varepsilon \log(1/\varepsilon)$, which is within a logarithmic factor of being sharp.

An important tool in our analysis is a new L^1 -functional inequality (Proposition 3.4). It arises as a limit case of a family of L^p -functional inequalities introduced by Meyer [37] and more recently revisited in [20, 36] in the Gaussian setting for p > 1. As we shall see, the L^1 -inequality fails in the Gaussian setting, but holds for RCD(N-1,N) spaces. Another new result of independent interest (Theorem 3.5) is a general criterion for proving that an eigenfunction has a distribution close to a Beta distribution, which is a variant of a result of E. Meckes [35] for the Gaussian setting.

As noted in [8, 32, 41, 46], the definition of the Bakry–Emery condition also makes sense for *negative* values of N. Sharp functional inequalities and model spaces in this setting were studied for example in [38]. Rigidity

in the smooth setting was studied in [34]. Under this curvature condition, the manifold may have infinite volume, so we shall require finite volume as an extra condition. Then the one-dimensional model space is \mathbb{R} endowed with a non-constant metric and a generalized Cauchy distribution [21]. We shall derive a stability estimate similar to that of the $\mathrm{RCD}(1,\infty)$ case, under an additional integrability condition on the eigenfunction, as well as assuming smoothness. To do so, we derive a version of Stein's lemma for one-dimensional generalized Cauchy distributions, of independent interest. We shall also give a Cauchy counterpart to E. Meckes' theorem on Gaussianity of eigenfunctions we previously mentioned.

The sequel is as follows: in Section 2, we shall present background results on RCD spaces, and a description of the model spaces for both positive and negative values of the dimension parameter. Section 3 shall contain the proof of our main Theorem, while Sections 4 and 5 shall respectively contain the results on the infinite-dimensional case and the negative-dimensional case.

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2. Preliminaries on the setting and curvature-dimension condition

We briefly explain in next sections what we call *smooth space* and *metric measure space*.

2.1. Some generalities on smooth setting

We briefly describe here the smooth setting, a complete description can be found in [7, Section 3.2].

The smooth setting that we consider in this paper, other possibilities are available in the literature, is a smooth weighted manifold. Let (M,g) be a (\mathcal{C}^{∞}) connected and complete d-dimensional Riemannian manifold $(d \geq 1)$. Let W be a smooth function on M and let $\mu = e^{-W}$ Vol be the

reference measure where Vol is the Riemannian measure. We assume in this article that μ is a probability measure and it is the case under the curvature-dimension condition used here.

The generator is described on smooth functions $f \in \mathcal{C}_c^{\infty}(M)$ by

$$\mathcal{L}(f) = \Delta_g f - \Gamma(W, f),$$

where Δ_g is the Laplace–Beltrami operator and Γ is the associated carré du champ operator. The Markov semigroup with generator \mathcal{L} is noted $(P_t)_{t\geqslant 0}$. For any smooth functions f, h, we have

$$\Gamma(f,h) = \langle \nabla f, \nabla h \rangle_g = \nabla f \cdot \nabla h,$$

that is scalar product associated with the metric q.

The weighted manifold (M,g) associated with the measure μ satisfies a CD(K,N) condition whenever the inequality (1.1) is satisfied for all smooth function f. Following for instance [7, Section C.6], if $N \in \mathbb{R} \setminus [0,d]$ then the curvature-dimension condition CD(K,N) is equivalent to the following inequality on tensors,

(2.1)
$$\operatorname{Ric}_{g} - \nabla \nabla^{g} W \geqslant Kg + \frac{1}{N - d} \nabla W \otimes \nabla W,$$

where Ric_g is the Ricci tensor of (M,g) and $\nabla \nabla^g W$ is the Hessian of W with respect to the metric g.

2.2. Some generalities on metric measure spaces

We consider a complete, separable, metric measure space (M, d, μ) , and we shall always assume that μ is a probability measure. We can define the Cheeger energy of an L^2 -function f as

$$\operatorname{Ch}(f) := \frac{1}{2} \inf_{(f_i)_{i \in \mathbb{N}}} \liminf_{i \to \infty} \int \left(\operatorname{Lip} f_i \right)^2 d\mu,$$

where the infimum runs over all sequences of locally-lipschitz functions converging to f in L^2 , and Lip f(x) is the local lipschitz constant at x. If an L^2 -function f has finite Cheeger energy, then there exists a minimal weak upper gradient, which we shall denote as $\Gamma(f)^{1/2}$, such that

$$Ch(f) = \frac{1}{2} \int \Gamma(f) d\mu.$$

The Sobolev space $W^{1,2}(M,d,\mu)$ is the space of $L^2(\mu)$ -functions with finite Cheeger energy. We refer to [1, 3] for more about these notions. For smooth functions on a Riemanian manifold, $\Gamma(f)$ coincides with $|\nabla f|^2$.

The space is said to be infinitesimally Hilbertian if the Cheeger energy is quadratic, that is

$$Ch(f+g) + Ch(f-g) = 2Ch(f) + Ch(g),$$

for all $f, g \in W^{1,2}$. Smooth manifolds are of course infinitesimally Hilbertian spaces, but there are examples of spaces, such as Finsler spaces, that do not satisfy this condition.

If the space is infinitesimally Hilbertian, we can define the scalar product $\Gamma(f,g) = \langle \nabla f, \nabla g \rangle \in L^1$ of two elements of $W^{1,2}$ by polarization of Γ , as well as the Dirichlet form $\mathcal{E}(f,g) = \int \Gamma(f,g) \, \mathrm{d}\mu$. The analogue of the Laplace operator for the space (M,d,μ) is then the operator $\mathcal{L}: D(\mathcal{L}) \longrightarrow L^2$ such that

$$\mathcal{E}(f,g) = -\int g(\mathcal{L}f) \,\mathrm{d}\mu.$$

The domain $D(\mathcal{L})$ of the operator is dense in L^2 . The associated Markov semigroup is also noted $(P_t)_{t\geqslant 0}$. We refer to [2] for background about this construction. We can then define RCD spaces as follows:

DEFINITION 2.1 (RCD(K,N) spaces). — A complete, separable, metric-measure space (M,d,μ) is said to be an RCD(K,N) space with $K \in \mathbb{R}$ and $N \geqslant 1$ if it is infinitesimally Hilbertian, it satisfies the Sobolev-to-Lipschitz property, and if it satisfies the Bochner inequality in a weak form: for any $f \in D(\mathcal{L})$ with $\mathcal{L}f \in W^{1,2}(M,d,\mu)$ and $g \in D(\mathcal{L}) \cap L^{\infty}(\mu)$ with $g \geqslant 0$ and $\mathcal{L}g \in L^{\infty}$ we have

$$(2.2) \ \frac{1}{2} \int \mathcal{L}g\Gamma(f) \,\mathrm{d}\mu - \int g\Gamma(f,\mathcal{L}f) \,\mathrm{d}\mu \geqslant K \int g\Gamma(f) \,\mathrm{d}\mu + \frac{1}{N} \int g(\mathcal{L}f)^2 \,\mathrm{d}\mu.$$

The main issue differentiating this weak formulation of the pointwise Bochner inequality (1.1) is that $\mathcal{L}\Gamma(f)$ might not be well-defined. Examples of RCD spaces include classical smooth manifolds satisfying Ricci curvature bounds, but also their possible Gromov–Hausdorff limits, as well as certain stratified spaces that do not arise as limits of smooth manifolds [11]. When N > 1, this definition is equivalent to the Lott–Sturm–Villani definition of Ricci curvature lower bounds, up to the extra assumption of linearity of the heat flow.

The Sobolev-to-Lipschitz property used in the definition is that for any function f with finite Cheeger energy, if $\Gamma(f) \leq 1$ μ -a.e., then it admits a 1-lipschitz representative. Honda gave examples of spaces satisfying the weak Bochner formula and not this assumption [24].

We will also be interested in the case where the dimension parameter is negative. The Bochner formula still makes sense for negative N, and there are interesting examples, such as the Cauchy distribution on \mathbb{R} . For this situation, we shall restrict ourselves to the smooth setting, since non-smooth issues such as the connection between the several possible definitions, are yet unclear.

The now classical bound on the spectral gap is the following, proved for example in [18, Theorem 4.22] in the case of a positive dimension parameter N in the RCD setting, and simultaneously in [41, Corollary 4.2] and in [31, Theorem 3.1] in the case of a negative dimension parameter N:

THEOREM 2.2 (Spectral gap of \mathcal{L}). — Let (M, d, μ) a RCD (ρ, N) metric measure space with $\rho > 0$ and $N \in \mathbb{R} \setminus [0, 1]$, the first eigenvalue $\lambda_1 > 0$ of $-\mathcal{L}$ satisfies

$$\lambda_1 \geqslant \frac{N\rho}{N-1}.$$

Moreover, existence of an eigenfunction associated with the spectral gap was proved in [22] under the $RCD(\rho, \infty)$ condition with $\rho > 0$.

2.3. Model spaces in dimension 1

We describe here the two main model spaces in dimension 1 used in the paper. The idea is to briefly describe how they are defined.

Let us consider φ , a smooth and positive function on an open interval $I \subset \mathbb{R}$. And let define the generator

$$\mathcal{L}f = \varphi f'' - (\beta - 1)\varphi' f',$$

for smooth function f. This generator has a reversible measure $\varphi^{-\beta} dx$, for any smooth and compactly supported function on I,

$$\int_{I} f \mathcal{L} g \varphi^{-\beta} \, \mathrm{d}x = -\int_{I} \varphi f' g' \varphi^{-\beta} \, \mathrm{d}x.$$

The carré du champ operator Γ , defined by

$$\Gamma(f) = \frac{1}{2}\mathcal{L}(f^2) - f\mathcal{L}f,$$

is given by $\Gamma(f) = \varphi \times (f')^2$. It says that we are working on the open interval I associated with the metric $1/\varphi$. From this new metric, the generator \mathcal{L} takes the form

$$\mathcal{L}f = \Delta_{\varphi}f - \Gamma(W, f),$$

where $\Delta_{\varphi} f = \varphi f'' + \frac{\varphi'}{2} f'$ is the Laplace–Beltrami operator on the one dimensional manifold $(I, 1/\varphi)$ and $W = (\beta - \frac{1}{2}) \log \varphi$.

Following [7, Section C.6], here in dimension 1, the operator \mathcal{L} satisfies the curvature-dimension condition $CD(\rho, N)$ (with $N \notin [0, 1]$) if and only if

(2.3)
$$\nabla \nabla^{\varphi} W - \frac{\rho}{\varphi} \geqslant \frac{(W')^2}{N-1},$$

where $\nabla \nabla^{\varphi} W = W'' + W' \frac{\varphi'}{2\varphi}$ is the Hessian of W with respect to the metric $1/\varphi$. Equation (2.3) becomes,

(2.4)
$$\left(\beta - \frac{1}{2}\right) \left[\varphi'' - \frac{\rho}{\beta - \frac{1}{2}} - \frac{\varphi'^2}{\varphi} \frac{N + 2(\beta - 1)}{2(N - 1)}\right] \geqslant 0.$$

The two main examples, used in the paper, are when there is equality in (2.4).

(1) Let I = (-1,1) and $\varphi = 1 - x^2$, $\rho = 1 - 2\beta$ and $N = 2(1 - \beta)$, $(\beta < 1/2)$. Then the generator

$$\mathcal{L}^+ f = (1 - x^2)f'' - Nxf'$$

is the so-called Jacobi operator and satisfies the curvature-dimension condition CD(N-1,N) with N>1. The carré du champ is $\Gamma(f)=(1-x^2)(f')^2$ and the reversible measure is a Beta distribution,

(2.5)
$$d\mu_N^+ = \frac{(1-x^2)^{N/2-1}}{Z^+} 1_{[-1,1]} dx,$$

where Z^+ is such that μ_N^+ it a probability measure. This measure is the so-called symmetrized Beta distribution with parameters (N/2, N/2).

The first non-trivial eigenvalue for the operator $-\mathcal{L}^+$ is N, with the eigenfunction f(x) = x. Moreover we have

$$\operatorname{Var}_{\mu_N^+}(f) = \int x^2 \, d\mu_N^+ - \left(\int x \, d\mu_N^+ \right)^2 = \int x^2 \, d\mu_N^+ = \frac{1}{N+1},$$

while
$$\int \Gamma(f) d\mu_N^+ = \frac{N}{N+1}$$
.

When $N \geqslant 2$ is a positive integer, μ_N^+ is the distribution of a coordinate on the N-dimensional unit sphere \mathbf{S}^N .

(2) Let now $I = \mathbb{R}$, $\varphi = 1 + x^2$, $\rho = 2\beta - 1$ and $N = 2(1 - \beta)$. We assume that $\beta > 3/2$, so that N < -1. For this model, we have

$$\mathcal{L}^- f = (1 + x^2)f'' + Nxf'$$

and $\Gamma(f) = (1 + x^2)f'^2$, and the reversible measure is a Cauchy type distribution (also called Student distribution),

(2.6)
$$d\mu_N^- = \frac{(1+x^2)^{N/2-1}}{Z^-} dx,$$

where Z^- is such that μ_N^- it is a probability measure. This model satisfies the curvature-dimension condition $\mathrm{CD}(1-N,N)$ with N<-1.

Again, the first non trivial eigenvalue for the operator \mathcal{L}^- is N with the eigenfunction is f(x) = x. We also have $\operatorname{Var}_{\mu_N^-}(f) = \frac{-1}{N+1}$ and $\int \Gamma(f) \, \mathrm{d}\mu_N^- = \frac{N}{N+1}$.

The assumption N < -1 is a necessary condition for the variance of f to exist.

Remark 2.3. — Of course, these models could be parametrized differently, without effect on the curvature or the spectral gap. In particular, our model for negative N is the same as the model space in [38], except that we parametrize it differently. It is important to notice that we chose to deal with these parametrizations of the two models (in dimension 1) so that the eigenfunction associated with the first eigenvalue is the identity function, and satisfies $\int \Gamma(f) d\mu_N^{\pm} = \frac{N}{N+1}$.

2.4. L^1 -Wasserstein distance

The W_1 distance, also called the L^1 -Wasserstein distance is the optimal transport distance in L^1 . For any μ, ν probability measure on a metric space (M, d),

$$W_1(\mu, \nu) = \inf \int \mathrm{d}(x, y) \, \mathrm{d}\pi(x, y),$$

where the infimum is running over all probability measures π on $M \times M$ which admit μ and ν as marginals. From the Kantorovich–Rubinstein's theorem (see [1] for instance) we have

$$W_1(\mu, \nu) = \sup_{\|g\|_{Li\nu} \leqslant 1} \left\{ \int g \, \mathrm{d}\mu - \int g \, \mathrm{d}\nu \right\},\,$$

and this is the definition used in this paper.

3. The positive and finite-dimensional case

Let N>1 be a real number and (M,d,μ) be a RCD(N-1,N) metric measure space as proposed in Definition 2.1. As is classical, we consider this case rather than more general RCD (ρ,N) spaces since we can normalize the value of ρ to N-1 by scaling the metric. From Theorem 2.2, its spectral gap satisfies $\lambda_1 \geqslant N$.

The next result proposes an estimate depending only on $(\lambda_1 - N)$, of the L^1 -Wasserstein distance between one direction of the measure μ and the reference measure μ_N^+ .

THEOREM 3.1 (The positive dimensional case). — Let (M, d, μ) be an RCD(N-1, N) metric measure space with N > 1 and generator \mathcal{L} . Let f be an eigenfunction of $-\mathcal{L}$ with eigenvalue $N + \varepsilon$, for $\varepsilon \in [0, 1)$, satisfying $\int \Gamma(f) d\mu = N/(N+1)$. Then

$$W_1(\mu \circ f^{-1}, \mu_N^+) \leqslant C\varepsilon,$$

where C > 0 is an explicit constant, depending only on N.

The proof of this result is postponed in Section 3.4. We first prove the two main ingredients: L^1 -estimates on eigenfunctions, and a criterion for comparing the distribution of an eigenfunction to a Beta distribution.

3.1. Estimates on the first eigenfunction

Our key technical estimate is the following:

LEMMA 3.2. — Let assume that \mathcal{L} satisfies the RCD(N-1,N) condition. Let f be an eigenfunction of $-\mathcal{L}$ with eigenvalue $N + \varepsilon$, for $\varepsilon \in [0,1)$ and satisfying $\int \Gamma(f) d\mu = N/(N+1)$. Then,

$$\|\Gamma(f) + (1+\varepsilon)f^2 - 1\|_1 \leqslant \varepsilon C$$

for some constant C depending only on N. The value

$$C = 4\left(2 + \frac{N+1}{N}\left(\frac{2}{N+1}\log 2 + \log\left(2 + \frac{2N}{(N-1)^2}\right)\right)\right) + \frac{N-1}{N(N+1)}$$

suffices.

The requirement that $\varepsilon < 1$ is for convenience, as it allows to simplify the writing of various bounds in the proof.

Estimates on $h = \Gamma(f) + (1 + \varepsilon)f^2$ are at the core of all of the results on rigidity and stability of sharp functional inequalities in the

RCD(N-1,N) setting. For rigidity, [29, Theorem 3.7] proves the $\varepsilon=0$ case of our lemma. In the smooth unweighted setting [9] uses L^p -estimates on Hessf+f (which is related to the gradient of $\Gamma(f)+f^2$), while [5] uses L^{∞} -estimates on $\Gamma(f)+f^2$. In the non-smooth setting, [13] establishes L^2 -estimates on Hessf+f along one-dimensional needles.

Unlike these previous work, we use a weaker norm to estimate h. One of the upsides is that it is easier to work with first order derivatives (instead of Hessians) in the non-smooth setting. But the main upside is that the quantitative bounds we derive are stronger. Indeed, one could use the self improvement in the Bakry–Emery–Bochner bound [45] to estimate Hessian-like quantities (this is the approach used in [10, 29] for example), but the quantitative bounds are of order $\sqrt{\varepsilon}$ instead of ε . Our use of an L^1 -norm leads instead to a sharp quantitative bound.

Our proof of Lemma 3.2 is based on the following result.

LEMMA 3.3. — Let assume that \mathcal{L} satisfies the RCD(N-1, N) condition with N > 1. Let f be an eigenfunction of $-\mathcal{L}$, with eigenvalue $N + \varepsilon$ for some $\varepsilon \in [0, 1)$. Then for $h = \Gamma(f) + (1 + \varepsilon)f^2$ we have

$$\|\mathcal{L}P_s h\|_1 \leqslant 4N\varepsilon \|f^2\|_1$$

where P_s is the semigroup generated from \mathcal{L} . In particular, in the smooth setting,

$$\|\mathcal{L}h\|_1 \leqslant 4N\varepsilon \|f^2\|_1.$$

The use of the semigroup P_s in the non-smooth setting is to avoid giving a meaning to $\mathcal{L}\Gamma(f)$. To simplify the exposition, we shall first prove Lemma 3.3 in the smooth setting, (assuming the Bochner inequality (1.1)) and then in the general RCD setting.

Proof of Lemma 3.3 in the smooth setting. — Applying the smooth CD(N-1,N) condition (1.1) to the eigenfunction f (recall that f is a smooth function since we are working in a smooth setting), we have

$$\frac{1}{2}\mathcal{L}\Gamma(f) + (1+\varepsilon)\Gamma(f) - \frac{(N+\varepsilon)^2}{N}f^2 \geqslant 0.$$

Moreover, since $2\Gamma(f) = \mathcal{L}(f^2) + 2(N+\varepsilon)f^2$, we have

$$\mathcal{L}(\Gamma(f) + (1+\varepsilon)f^2) + 2(\varepsilon(N-1) + \varepsilon^2(1-1/N))f^2 \ge 0.$$

In particular,

$$(\mathcal{L}h)_{-} \leqslant 2\varepsilon(N-1)\left(1+\frac{\varepsilon}{N}\right)f^{2} \leqslant 2\varepsilon Nf^{2},$$

using $\varepsilon < 1$ (which is only used here to make notations less cluttered) and N > 1. Since $\mathcal{L}h$ has zero average with respect to μ , we get

$$\|\mathcal{L}h\|_1 = 2\|(\mathcal{L}h)_-\|_1 \leqslant 4\varepsilon N\|f^2\|_1.$$

Which concludes the proof.

Proof of Lemma 3.3 in the general RCD case. — As mentioned previously, to follow the above scheme in the general RCD setting, we encounter the problem of giving a meaning to $\mathcal{L}\Gamma(f)$. We shall exploit the particular structure of f as an eigenfunction to bypass this issue via a regularization procedure.

The weak Bochner inequality (2.2) applied to the eigenfunction f of $-\mathcal{L}$ with eigenvalue $N + \varepsilon$ takes the form:

$$\frac{1}{2} \int \mathcal{L}g\Gamma(f) \, \mathrm{d}\mu + (N+\varepsilon) \int g\Gamma(f) \, \mathrm{d}\mu$$

$$\geqslant (N-1) \int g\Gamma(f) \, \mathrm{d}\mu + \frac{(N+\varepsilon)^2}{N} \int gf^2 \, \mathrm{d}\mu,$$

for any test function g such that $g, \mathcal{L}g \in L^{\infty}(\mu)$.

If we consider test functions of the form $P_s g$ for s > 0 and $g, \mathcal{L}g \in L^{\infty}(\mu)$, since \mathcal{L} and P_s commute and $\mathcal{L}P_s\Gamma(f)$ is well defined we have

(3.1)
$$\frac{1}{2} \int g(\mathcal{L}P_s\Gamma(f)) d\mu$$

$$\geqslant -(\varepsilon+1) \int gP_s\Gamma(f) d\mu + \frac{(N+\varepsilon)^2}{N} \int gP_s(f^2) d\mu.$$

We can now remove the restriction that $\mathcal{L}g$ is $L^{\infty}(\mu)$ by approximating a $g \in L^{\infty}(\mu)$ with

$$S_{\varepsilon}g := \int_0^\infty P_{\varepsilon r} g \kappa(r) \, \mathrm{d}r,$$

where κ is a smooth nonnegative function compactly supported in $(0, \infty)$ with $\int \kappa(r)dr = 1$. One can check that

$$\mathcal{L}S_{\varepsilon}g = -\frac{1}{\varepsilon} \int_0^{\infty} P_{\varepsilon r} g \kappa'(r) \, \mathrm{d}r$$

is indeed $L^{\infty}(\mu)$, and that $S_{\varepsilon}g$ converges to g in $L^{2}(\mu)$ when $\varepsilon \to 0$. Therefore (3.1) holds for $g \in L^{\infty}(\mu)$. But since $P_{s}\Gamma(f)$ is a well-defined function, we deduce the pointwise inequality

$$\frac{1}{2}\mathcal{L}P_s\Gamma(f)\geqslant -(\varepsilon+1)P_s\Gamma(f)+\frac{(\varepsilon+N)^2}{N}P_s(f^2),\quad s>0.$$

Since $\Gamma(f) = \frac{1}{2}\mathcal{L}f^2 + (\varepsilon + N)f^2$, we get as in the smooth setting, for $h = \Gamma(f) + (1 + \varepsilon)f^2$,

$$(\mathcal{L}P_s h)_- \leqslant 2\varepsilon N P_s(f^2).$$

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Hence we can deduce that $\|\mathcal{L}P_s h\|_1 \leq 4N\varepsilon \|f\|_2^2$.

Since the kernel of \mathcal{L} is the set of constants, we expect that h is concentrated around its average. A tool for proving this is provided by the following result.

PROPOSITION 3.4. — Let assume that \mathcal{L} satisfies the condition RCD(N-1,N) with N>1, then for any function $g\in D(\mathcal{L})$ with $\int g\,\mathrm{d}\mu=0$, we have

$$||g||_1 \le \left(2 + \frac{N+1}{N} \left(\frac{2}{N+1} \log 2 + \log \left(2 + \frac{2N}{(N-1)^2}\right)\right)\right) ||\mathcal{L}g||_1.$$

Proof. — We have for any t > 0,

$$||g||_1 \leq ||g - P_t g||_1 + ||P_t g||_1.$$

For any bounded test function u, we have

$$\int (g - P_t g) u \, d\mu = \int_0^t \int (-\mathcal{L} P_s g) u \, d\mu \, ds$$
$$= \int_0^t \int (-\mathcal{L} g) (P_s u) \, d\mu \, ds \leqslant t \|u\|_{\infty} \|\mathcal{L} g\|_1.$$

Taking the supremum over all u with $||u||_{\infty} \leq 1$, we get

$$||g - P_t g||_1 \leqslant t ||\mathcal{L}g||_1,$$

hence

(3.2)
$$||g||_1 \leq (t+1)||\mathcal{L}g||_1 + ||P_{t+1}g||_1.$$

Since N > 1, the operator \mathcal{L} satisfies the weaker condition CD(N-1, N+1). From this remark, \mathcal{L} satisfies a Sobolev inequality (cf. [26] for the smooth setting and extended to the RCD setting in [43]),

$$||f||_{2\frac{N+1}{N-1}}^2 \le ||f||_2^2 + B \int \Gamma(f) \,\mathrm{d}\mu,$$

with $B = \frac{4N}{(N+1)(N-1)^2}$ and for functions such that terms are well defined. And then, the ultracontractive bound

(3.3)
$$||P_t g||_{\infty} \leqslant C t^{-\frac{N+1}{2}} ||g||_1, \quad t \leqslant 1$$

with $C = \left(2 + \frac{2N}{(N-1)^2}\right)^{\frac{N+1}{2}}$, see [7, Theorem 6.3.1 and Remark 6.3.2]. Using this bound for t=1 and the spectral gap, we have

$$||P_{t+1}g||_1 \le ||P_{t+1}g||_2$$

$$\le e^{-Nt} ||P_1g||_2$$

$$\le e^{-Nt} \left(2 + \frac{2N}{(N-1)^2}\right)^{\frac{N+1}{2}} ||g||_1.$$

Taking

$$t = \frac{N+1}{2N} \left(\frac{2}{N+1} \log 2 + \log \left(2 + \frac{2N}{(N-1)^2} \right) \right)$$

and using inequality (3.2), we get

which is the inequality desired.

$$||g||_1 \le (t+1)||\mathcal{L}g||_1 + \frac{||g||_1}{2}.$$

Hence

$$||g||_1 \le 2(t+1)||\mathcal{L}g||_1$$

$$= \left(2 + \frac{N+1}{N} \left(\frac{2}{N+1} \log 2 + \log \left(2 + \frac{2N}{(N-1)^2}\right)\right)\right) ||\mathcal{L}g||_1,$$

Proof of Lemma 3.2. — Let apply Proposition 3.4 to

$$P_s\Gamma(f) + (1+\varepsilon)P_s(f^2) - \frac{N}{N+1}\left(1+\frac{1+\varepsilon}{N+\varepsilon}\right)$$

for some s > 0, which has zero average since $\int f^2 d\mu = (N + \varepsilon)^{-1} \int \Gamma(f) d\mu$. We obtain,

$$\left\| P_s \Gamma(f) + (1+\varepsilon) P_s(f^2) - \frac{N}{N+1} \left(1 + \frac{1+\varepsilon}{N+\varepsilon} \right) \right\|_1 \\ \leqslant C \| \mathcal{L} \left[P_s \Gamma(f) + (1+\varepsilon) P_s(f^2) \right] \|_1,$$

where C is given by Proposition 3.4. And then, using Lemma 3.3,

$$\left\| P_s \Gamma(f) + (1+\varepsilon) P_s(f^2) - \frac{N}{N+1} \left(1 + \frac{1+\varepsilon}{N+\varepsilon} \right) \right\|_1 \leqslant 4CN\varepsilon \|f\|_2 \leqslant 4C\varepsilon.$$

Finally,

$$\|P_s\Gamma(f) + (1+\varepsilon)P_s(f^2) - 1\|_1 \leqslant \varepsilon \left(4C + \frac{N-1}{N(N+1)}\right),$$

from the triangle inequality. We let s go to zero to conclude the proof. \square

3.2. Approximate Beta distribution for eigenfunctions

The last main ingredient of the proof is a result stating that (normalized) eigenfunctions with eigenvalue close to N and such that $\Gamma(f) + f^2$ is close to a constant approximately follow a symmetrized Beta distribution. The result, of independent interest, is the following:

THEOREM 3.5. — Let f be an eigenfunction of a diffusion operator $-\mathcal{L}$ with eigenvalue λ and invariant probability measure μ , and let $\nu = \mu \circ f^{-1}$ be the pushforward of μ by f. Then

$$W_1(\nu, \text{Beta}(N/2, N/2)) \leqslant \left(\frac{N^2}{2} + \frac{5N}{2} + \frac{7}{2}\right) \|\Gamma(f) + f^2 - 1\|_1 + \frac{2|N - \lambda|}{N} \|f\|_2,$$

where $\text{Beta}(N/2, N/2) = \mu_N^+$ (defined in Section 2.3).

Note that this result is only interesting if λ is close to N, and if f has been normalized so that $\int (\Gamma(f) + f^2) d\mu$ is close to one.

This result is a variant of a result of E. Meckes [35, Theorem 1], who proved that eigenfunctions whose gradient has small variance are close to normal. As in Meckes' work, the proof will mostly be an application of Stein's method.

3.2.1. Stein's method for Beta distributions

To prove Theorem 3.5, we shall rely on a variant of Stein's method for Beta distributions. Stein's method is a set of techniques, pioneered in [47, 48], for bounding distances between probability measures via such integration-by-parts formulas. We refer to [14, 44] for recent introductions and surveys of this field.

The following variant of Stein's lemma was proven in [17, 23], for non-symmetric Beta distributions on [0, 1] (and the symmetrized case is an immediate consequence). The values of the constants stated here are slightly worse than those stated in [17, Proposition 4.2], and used here for ease of writing.

Theorem 3.6 ([17, 23]). — Let N > 1, the W_1 distance between a probability measure ν supported on [-1,1] to a Beta(N/2,N/2) distribution can be estimated by

(3.4)
$$W_1(\nu, \text{Beta}(N/2, N/2)) \leq \frac{1}{2} \sup \left\{ \int \left[(1 - x^2)g'(x) + Nxg(x) \right] d\nu \right\},$$

where the supremum is running over all smooth function g on $\mathbb R$ such that

$$||g||_{\infty} \leqslant \frac{2}{N}$$
 and $||g'||_{\infty} \leqslant (2+N)$.

3.3. Proof of Theorem 3.5

We want to apply Theorem 3.6 to the measure $\nu = \mu \circ f^{-1}$. We first derive an approximate integration by parts formula for the measure ν . Let g be a smooth test function with $||g||_{\infty} \leq 2/N$ and $||g'||_{\infty} \leq (2+N)$. Using the diffusion property (1.2) and the definition of ν , we have

$$\lambda \int x g(x) \, d\nu(x) = \lambda \int f g \circ f \, d\mu$$

$$= \int (-\mathcal{L}f) g \circ f \, d\mu$$

$$= \int \Gamma(f, g \circ f) \, d\mu$$

$$= \int g' \circ f \Gamma(f) \, d\mu.$$

So

$$\int \left[\lambda x g(x) - (1 - x^2) g'(x)\right] d\nu(x)$$

$$= \int g' \circ f\Gamma(f) d\mu - \int (1 - x^2) g'(x) d\nu$$

$$= \int \left[\Gamma(f) + f^2 - 1\right] g' \circ f d\mu \leqslant (2 + N) \|\Gamma(f) + f^2 - 1\|_1,$$

where we use the definition of $\nu = \mu \circ f^{-1}$. Moreover

$$\left| \int x g(x) \, \mathrm{d}\nu(x) \right| = \left| \int f g(f) \, \mathrm{d}\mu \right| \leqslant \|g\|_{\infty} \|f\|_{L^2(\mu)} \leqslant \frac{2}{N} \|f\|_2.$$

Therefore, we get

$$(3.5) \left| \int \left[Nxg(x) - \left(1 - x^2\right)g'(x) \right] d\nu(x) \right|$$

$$\leq \left| \int \left[\lambda xg(x) - \left(1 - x^2\right)g'(x) \right] d\nu(x) \right| + |N - \lambda| \left| \int xg(x) d\nu(x) \right|$$

$$\leq (2 + N) \|\Gamma(f) + f^2 - 1\|_1 + 2 \frac{|N - \lambda|}{N} \|f\|_2.$$

To apply Theorem 3.6, we still need to consider a measure supported on [-1,1]. To do so, we introduce $\tilde{\nu}$ the pushforward of μ by $\phi \circ f$, with ϕ a cutoff function, that is $\phi(x) = x$ on [-1,1], $\phi(x) = 1$ on $[1,+\infty[$, and

$$\phi(x) = -1$$
 on $]-\infty, -1]$. Note that

$$\int |f - \phi(f)| \, \mathrm{d}\mu = \int (|f| - 1) \mathbf{1}_{|f| \ge 1} \, \mathrm{d}\mu$$

$$\leqslant \frac{1}{2} \int (f^2 - 1) \mathbf{1}_{|f| \ge 1} \, \mathrm{d}\mu$$

$$= \frac{1}{2} \int (f^2 - 1)_+ \, \mathrm{d}\mu$$

$$\leqslant \frac{1}{2} \int (\Gamma(f) + f^2 - 1)_+ \, \mathrm{d}\mu$$

$$\leqslant \frac{1}{2} \|\Gamma(f) + f^2 - 1\|_1.$$

Therefore by a coupling argument

(3.7)
$$W_1(\widetilde{\nu}, \nu) \leqslant \int x \, d\nu(x) - \int x \, d\widetilde{\nu}(x)$$
$$\leqslant \int |f - \phi(f)| \, d\mu \leqslant \frac{1}{2} \|\Gamma(f) + f^2 - 1\|_1.$$

It is therefore enough to apply Theorem 3.6 to $\tilde{\nu}$. To do so, we shall show that it satisfies the same approximate integration by parts formula than ν , up to an error of order $\|\Gamma(f) + f^2 - 1\|_1$.

First we have,

$$\begin{split} \left| \int [fg(f) - \phi(f)g(\phi(f))] \, \mathrm{d}\mu \right| \\ & \leqslant \left| \int (f - \phi(f))g(f) \, \mathrm{d}\mu \right| + \left| \int \phi(f)(g(f) - g(\phi(f))) \, \mathrm{d}\mu \right| \\ & \leqslant \frac{2}{N} \|f - \phi(f)\|_1 + (2 + N) \|f - \phi(f)\|_1 \\ & \leqslant \left(1 + \frac{1}{N} + \frac{N}{2} \right) \|\Gamma(f) + f^2 - 1\|_1, \end{split}$$

since $\|\phi(f)\|_{\infty} \leq 1$. Secondly,

$$\left| \int \left[(1 - f^2) g'(f) - (1 - \phi(f)^2) g'(\phi(f)) \right] d\mu \right|$$

$$\leq \left| \int (\phi(f)^2 - f^2) g'(f) d\mu \right| + \left| \int (1 - \phi(f)^2) (g'(f) - g'(\phi(f))) d\mu \right|$$

$$\leq (2 + N) \int (f^2 - 1)_+ d\mu + 0$$

$$\leq (2 + N) \|\Gamma(f) + f^2 - 1\|_1.$$

where we used that

$$\int (1 - \phi(f)^2)(g'(f) - g'(\phi(f))) d\mu = 0$$

(this is clear by computing separately the integral on $|f| \leq 1$ and on $|f| \geq 1$), and we again used (3.6) to see that

$$\int |\phi(f)^2 - f^2| \, \mathrm{d}\mu = \int_{|f| \ge 1} (f^2 - 1) \, \mathrm{d}\mu \le \|\Gamma(f) + f^2 - 1\|_1.$$

Hence from the definition of $\tilde{\nu}$, we have

$$\left| \int (Nxg(x) - (1 - x^2)g'(x)) \, d\tilde{\nu}(x) \right|$$

$$= \left| \int (N\phi(f)g(\phi(f)) - (1 - \phi(f)^2)g'(\phi(f))) \, d\mu \right|$$

$$\leq N \left| \int [\phi(f)g(\phi(f)) - fg(f))] \, d\mu \right|$$

$$+ \left| \int [(1 - f^2)g'(f) - (1 - \phi(f)^2)g'(\phi(f))] \, d\mu \right|$$

$$+ \left| \int [Nxg(x) - (1 - x^2)g'(x)] \, d\nu(x) \right|.$$

We can deduce from the previous estimates and (3.5),

$$\left| \int Nxg(x) - (1 - x^2)g'(x) \, d\tilde{\nu} \right|$$

$$\leq \left(\frac{N^2}{2} + \frac{5N}{2} + 3 \right) \|\Gamma(f) + f^2 - 1\|_1 + 2 \frac{|N - \lambda|}{N} \|f\|_2.$$

At the end, from Theorem 3.6 to $\tilde{\nu}$ and the estimate (3.7),

$$\begin{split} W_1(\nu, \text{Beta}(N/2, N/2)) \\ &\leqslant W_1(\widetilde{\nu}, \nu) + W_1(\widetilde{\nu}, \text{Beta}(N/2, N/2)) \\ &\leqslant \left(\frac{N^2}{2} + \frac{5N}{2} + \frac{7}{2}\right) \|\Gamma(f) + f^2 - 1\|_1 + \frac{2|N - \lambda|}{N} \|f\|_2, \end{split}$$

which is the inequality desired.

3.4. Proof of Theorem 3.1

We can now straightforwardly combine Theorem 3.5 and Lemma 3.2 to conclude.

Note that under our normalization $\int \Gamma(f) d\mu = N/(N+1)$, we have

$$||f^2||_1 \leqslant \frac{1}{N+1}$$

and hence

$$\|\Gamma(f) + f^2 - 1\|_1 \le \varepsilon \|f^2\|_1 + \|\Gamma(f) + (1+\varepsilon)f^2 - 1\|_1 \le C\varepsilon$$

for some explicit constant C > 0 depending only on N.

4. Stability of the spectral gap for $RCD(1, \infty)$ spaces

In this section, we are working with (M, d, μ) a RCD $(1, \infty)$ metric measure space with generator \mathcal{L} and unit mass.

Arguing as in the finite-dimensional case, we can use the Bochner formula to get the following estimate on the gradient of a normalized eigenfunction f:

(4.1)
$$\|\mathcal{L}P_s\Gamma(f)\|_1 \leqslant C\varepsilon, \quad s > 0.$$

However, unlike RCD(N-1, N) spaces, in this situation we do not have ultracontractive estimates on the semigroup (inequality (3.3)) to prove an L^1 -inequality as in Lemma 3.2. Indeed, as we shall see in Proposition 4.4 below, that inequality fails for the Gauss space. Therefore, we must rely on a weaker functional inequality:

LEMMA 4.1. — Let assume that the spectral gap of $-\mathcal{L}$ is greater than 1. Let g such that $\int g \, d\mu = 0$ and $g \in L^p(\mu)$ for some p > 1. Then

$$\|g\|_1 \leqslant C \|\mathcal{L}g\|_1 \left(1 + \log \left(\max \left(\frac{\|g\|_p}{\|\mathcal{L}g\|_1}, 1\right)\right)\right),$$

for some constant C depending only on p.

Note that this lemma does not involve any assumption on curvature. We will actually be interested in applying it when $||g||_p$ is bounded and $||\mathcal{L}g||_1$ is small.

Proof. — As before, using inequality (3.2), we obtain for $t \ge 0$,

$$||g||_1 \le t||\mathcal{L}g||_1 + ||P_tg||_1 \le t||\mathcal{L}g||_1 + ||P_tg||_p.$$

According to [12, Theorem 1.6], we have for some constant C_p (depending only p),

$$||P_t g||_p \leqslant C_p \exp(-4(p-1)t/p^2)||g||_p.$$

Hence, we have

$$||g||_1 \le t||\mathcal{L}g||_1 + C_p \exp(-4(p-1)t/p^2)||g||_p.$$

Assume for simplicity that $C_p \ge 1$ and take

$$t = \frac{p^2}{4(p-1)} \log \left[\max \left(\frac{\|g\|_p}{\|\mathcal{L}g\|_1}; 1 \right) C_p \right],$$

in order to get

$$C_p \exp(-4(p-1)t/p^2) \leqslant \frac{\|\mathcal{L}g\|_1}{\|g\|_p}.$$

We get

$$||g||_1 \le ||\mathcal{L}g||_1 \left[1 + \frac{p^2}{4(p-1)} \log \left(C_p \max \left(\frac{||g||_p}{||\mathcal{L}g||_1}, 1 \right) \right) \right],$$

which is the inequality expected.

LEMMA 4.2. — Let assume that \mathcal{L} satisfies the RCD $(1, \infty)$ condition. And let f be an eigenfunction of $-\mathcal{L}$ with eigenvalue $1 + \varepsilon$, for $\varepsilon \in [0, 1]$ and satisfying $\int \Gamma(f) d\mu = 1$. Then,

for some numerical C.

Proof. — Following the same proof as Lemma 3.3 in the general RCD setting, we get

$$\|\mathcal{L}(P_s\Gamma(f) + \varepsilon P_s(f^2))\|_1 \le 4\varepsilon \|f^2\|_1 \le 4\varepsilon,$$

since $\int f^2 d\mu = (1+\varepsilon)^{-1} \leq 1$. If $h = P_s(\Gamma(f) + \varepsilon f^2)$, we have $\int h d\mu = (1+2\varepsilon)(1+\varepsilon)^{-1}$ and then

$$\left\| h - \frac{1+2\varepsilon}{1+\varepsilon} \right\|_{1} \leqslant C \|\mathcal{L}h\|_{1} \left(1 + \log \left(\max \left(\frac{\|h\|_{p}}{\|\mathcal{L}h\|_{1}}, 1 \right) \right) \right)$$
$$\leqslant 4C\varepsilon \left(1 + \log \left(\max \left(\frac{\|h\|_{p}}{4\varepsilon}, 1 \right) \right) \right).$$

Hence

$$\begin{split} \|h-1\|_1 &\leqslant \left\|h - \frac{1+2\varepsilon}{1+\varepsilon}\right\|_1 + \frac{1+2\varepsilon}{1+\varepsilon} - 1 \\ &\leqslant 4C\varepsilon \left(1 + \log\left(\max\left(\frac{\|h\|_p}{4\varepsilon}, 1\right)\right)\right) + \varepsilon. \end{split}$$

From [10, Proposition 3.2] (and the $RCD(1, \infty)$ condition) we have,

$$||h||_p \le ||P_s\Gamma(f)||_p + ||P_sf||_{2p} \le ||\Gamma(f)||_p + ||f||_{2p}$$

$$\le (8p - 4)^{(1+\varepsilon)/2} ||f^2||_1 + (p - 1)^{(1+\varepsilon)/2} ||f^2||_1 \le D$$

for some other constant D > 0, which is the last estimate to prove (4.2). \square

Lemma 4.2 improves on a particular case of [10, Lemma 3.3], and is enough to improve [10, Theorem 1.3] into:

THEOREM 4.3. — Let (M, d, μ) a RCD $(1, \infty)$ metric measure space with generator \mathcal{L} . Let f is an eigenfunction of the $-\mathcal{L}$ with eigenvalue $1 + \varepsilon$ $(\varepsilon \in [0, 1])$ and satisfying $\int \Gamma(f) d\mu = 1$, then

$$W_1(\mu \circ f^{-1}, \gamma) \leqslant C\varepsilon \log(2/\varepsilon),$$

for some numerical constant C > 0 and γ is the standard Gaussian measure.

Proof. — We follow the same line of arguments as in [10] (and the implementation of Stein's method is essentially the same as the argument in [35]). From Stein's lemma for the one-dimensional Gaussian distribution (see for example [10, Lemma 3.5]) we have,

$$W_1(\mu \circ f^{-1}, \gamma) \le \sup \left\{ \int (g'(x) - g(x)x) \, \mathrm{d}\mu \circ f^{-1}(x), \|g'\|_{\infty} \le 4 \right\}.$$

Similarly we have $(1+\varepsilon)\int xg(x)\,\mathrm{d}\mu\circ f^{-1}(x)=\int g'\circ f\Gamma(f)\,\mathrm{d}\mu$, then

$$\int (g'(x) - g(x)x) d\mu \circ f^{-1}(x)$$

$$= \frac{1}{1+\varepsilon} \int \left[1 + \varepsilon - \Gamma(f) - \varepsilon f^2\right] g' \circ f d\mu - \frac{\varepsilon}{1+\varepsilon} \int f^2 g' \circ f d\mu.$$

That is for g satisfying $||g'||_{\infty} \leq 4$,

$$\int (g'(x) - g(x)x) d\mu \circ f^{-1}(x) \leqslant \frac{4}{1+\varepsilon} \|\Gamma(f) + \varepsilon f^2 - 1 - \varepsilon\|_1 + 4\varepsilon.$$

From the estimate (4.2) we get

$$\int (g'(x) - g(x)x) \, d\mu \circ f^{-1}(x) \leqslant C\varepsilon \log(2/\varepsilon),$$

for some numerical constant C, which concludes the proof.

We now show that indeed the L^1 -estimate of Lemma 3.2 may fail in the $\mathrm{RCD}(1,\infty)$ setting.

Proposition 4.4. — The inequality

$$||f||_1 \leqslant C||\mathcal{L}f||_1$$

for some constant C > 0 and every centered functions f fails in the Gauss space $(\mathbb{R}, |\cdot|, \gamma)$.

The counterexample in the proof below is inspired by Naor and Schechtman's proof [39] that an analogous inequality on the hypercube $\{0,1\}^d$ cannot hold with a constant that is uniform in d. We thank Alexandros Eskenazis for pointing out this reference to us.

Proof. — Let \mathcal{L} be the one-dimensional Ornstein–Uhlenbeck operator $\mathcal{L}f = f'' - xf'$. We shall exhibit a family of functions f_r such that the ratio

$$||f_r||_1/||\mathcal{L}f_r||_1$$

is unbounded. Let f_r be the centered solution (in $L^2(\gamma)$) to the Poisson equation

$$\mathcal{L}f_r = -\mathbf{1}_{(-\infty, -r]} + \mathbf{1}_{[r, +\infty)}.$$

Since the source term is antisymmetric, we are looking for an antisymmetric solution (which will then be centered). We can check that f'_r is given on \mathbb{R}_+ by the formula

$$f'_r(x) = \begin{cases} -\sqrt{2\pi} (1 - \varphi(r)) e^{x^2/2} & \text{if } 0 \le x \le r; \\ -\sqrt{2\pi} (1 - \varphi(x)) e^{x^2/2} & \text{if } x > r, \end{cases}$$

and extended by symmetry to \mathbb{R} . Here φ stands for the Gaussian cumulative distribution function $\varphi(x) = \int_{-\infty}^{x} (2\pi)^{-1/2} \exp(-t^2/2) dt$. Then for $x \in [0, r]$ we have

$$f_r(x) = -\sqrt{2\pi}(1 - \varphi(r)) \int_0^x e^{t^2/2} dt.$$

It is easy to check that for $x \in [1, r]$ we have a lower bound of the form

$$f_r(x) \geqslant C(1 - \varphi(r)) \frac{e^{x^2/2}}{x}, C > 0.$$

Therefore, for r > 1,

$$\int |f_r| \, d\gamma \geqslant C(1 - \varphi(r)) \int_{\sqrt{r}}^r \frac{e^{x^2/2}}{x} \, d\gamma = \frac{C}{\sqrt{2\pi}} (1 - \varphi(r)) \int_{\sqrt{r}}^r x^{-1} \, dx$$
$$= \frac{C}{2\sqrt{2\pi}} (1 - \varphi(r)) \log r.$$

On the other hand

$$\|\mathcal{L}f_r\|_1 = \int |\mathcal{L}f_r| \,\mathrm{d}\gamma = 2(1 - \varphi(r)).$$

Hence $||f_r||_1/||\mathcal{L}f_r||_1$ is unbounded as r goes to infinity, which concludes the proof.

5. The negative dimension case

We now consider a space satisfying the curvature-dimension condition with negative dimension parameter N. We refer to [32, 41, 46] for an introduction to the notion. Since the Bochner inequality in the non-smooth setting seems not to have been investigated yet in the literature (in particular, its possible equivalence with other non-smooth definitions studied in [41]), we shall restrict ourselves to the smooth setting, and assume the Bochner inequality (1.1) holds in a strong sense, as in [32]. As noted in [31], the spectral gap bound of Theorem 2.2 is sharp for $N \leq -1$, but ceases to be sharp when N is negative and |N| is small (see also [34] where spaces with negative effective dimension, positive curvature and infinite volume are studied). When $N \leq -1$, a particular example of a model satisfying a CD(1-N,N) condition is given by the generalized Cauchy distribution as presented in Section 2.3,

$$d\mu_N^- = \frac{(1+x^2)^{N/2-1}}{Z^-} dx$$

on \mathbb{R} , with generator

$$\mathcal{L}^{-}f = (1 + x^{2})f'' + Nxf'.$$

Our main result in this section is:

THEOREM 5.1 (The negative dimensional case). — Let (M,g) be a smooth Riemannian manifold associated with the probability measure μ , satisfying the curvature-dimension condition CD(1-N,N) with N<-1. We assume that the spectral gap λ_1 satisfies $\lambda_1 \leqslant -N + \varepsilon$ for some $\varepsilon \in (0,1)$, and let f an eigenfunction of \mathcal{L} with eigenvalue λ_1 such that $\int \Gamma(f) d\mu = N/(N+1)$.

Assume moreover that $\Gamma(f) \in L^{1+c}$ for some c > 0. There is a constant C > 0 depending only on N and c such that

$$W_1(\mu \circ f^{-1}, \mu_N^-) \leqslant C\varepsilon \log(2/\varepsilon).$$

This result shall be a direct consequence of combining Proposition 5.4 with Corollary 5.6 below.

5.1. Estimates on the first eigenfunction

As for Lemma 3.3 with the same proof, we deduce the following estimate for the Bochner formula:

LEMMA 5.2. — Let assume that \mathcal{L} satisfies the CD(1-N,N) condition. We assume that there exits f an eigenfunction of $-\mathcal{L}$ with eigenvalue $-N+\varepsilon$ with $\varepsilon \in [0,1]$ such that $f \in L^2(\mu)$. Then

(5.1)
$$\|\mathcal{L}(\Gamma(f) + (\varepsilon - 1)f^2)\|_1 \leqslant 4\varepsilon \frac{(1 - N)^2}{|N|} \|f^2\|_1.$$

Proof. — Applying the CD(1 - N, N) condition (1.1) to the eigenfunction f, we have

$$\frac{1}{2}\mathcal{L}\Gamma(f) + (\varepsilon - 1)\Gamma(f) - \frac{(N - \varepsilon)^2}{N}f^2 \geqslant 0.$$

Moreover, since $2\Gamma(f) = \mathcal{L}(f^2) + 2(\varepsilon - N)f^2$, we have

$$\frac{1}{2}\mathcal{L}(\Gamma(f)+(\varepsilon-1)f^2)+\varepsilon(\varepsilon-N)\frac{1-N}{|N|}f^2\geqslant 0.$$

In particular, for $h = \Gamma(f) + (\varepsilon - 1)f^2$,

$$(\mathcal{L}h)_{-} \leqslant 2\varepsilon(\varepsilon - N)\frac{1 - N}{|N|}f^{2} \leqslant -2\frac{\varepsilon}{N}(1 - N)^{2}f^{2},$$

using $\varepsilon < 1$. Since $\mathcal{L}h$ has zero average with respect to μ , we get

$$\|\mathcal{L}h\|_1 = 2\|(\mathcal{L}h)_-\|_1 \leqslant 4\frac{\varepsilon}{|N|} (1-N)^2 \|f^2\|_1,$$

which concludes the proof.

Once again, we do not have an ultracontractive estimate for the semi-group when the dimension parameter is negative, so we must rely on Lemma 4.1 to deduce an L^1 -estimate on $h = \Gamma(f) + (\varepsilon - 1)f^2$. This use of Lemma 4.1 would be justified if both $\Gamma(f) \in L^p(\mu)$ for some p > 1 and $f \in L^q(\mu)$ for some q > 2. The following lemma shows that only the integrability condition on $\Gamma(f)$ is actually required.

LEMMA 5.3. — Assume the couple (μ, Γ) satisfies a Poincaré inequality with constant C_P . If $\Gamma(g) \in L^{1+c}(\mu)$ for some c > 0 then

$$g \in L^{2(1+2c)/(1+c)}(\mu),$$

and

$$||g||_{2(1+2c)/(1+c)}^{2(1+2c)/(1+c)} \le ||g||_{2}^{2(1+2c)/(1+c)} + 4C_P ||\Gamma(g)||_{1+c} ||g||_{2}^{2c/(1+c)}.$$

Proof. — The Poincaré inequality insures that for all smooth function h,

$$\int h^2 d\mu - \left(\int h d\mu\right)^2 \leqslant C_P \int \Gamma(h) d\mu.$$

This inequality applied to g^p gives

$$||g||_{2p}^p - ||g||_p^{2p} \le p^2 C_P \int \Gamma(g) g^{2(p-1)} d\mu.$$

We then apply the Hölder inequality with exponents 1+c and (c+1)/c to get

$$||g||_{2p}^{2p} - ||g||_{p}^{2p} \le p^{2} C_{P} ||\Gamma(g)||_{1+c} \left(\int g^{2(p-1)(c+1)/c} d\mu \right)^{c/(c+1)}.$$

We then take p = 1 + c/(c+1), since p < 2 we have $||g||_p^{2p} \le ||g||_2^{2p}$, which concludes the proof.

PROPOSITION 5.4. — Assume that \mathcal{L} satisfies the $\mathrm{CD}(1-N,N)$ condition (with N<-1) and assume that \mathcal{L} admits an eigenfunction f with eigenvalue $-N+\varepsilon$ with $\varepsilon\leqslant 1$. We assume moreover that $f\in L^2(\mu)$, $\Gamma(f)\in L^{1+c}(\mu)$ for some c>0, and $\int \Gamma(f)\,\mathrm{d}\mu=\frac{N}{N+1}$. Then

$$\|\Gamma(f) - f^2 - 1\|_1 \leqslant C\varepsilon \log(2/\varepsilon),$$

for a numerical constant C that only depends on c and N.

Proof. — The proof is a straightforward combination of Lemmas 4.1, 5.2 and 5.3. Note that with the normalization of $\int \Gamma(f) d\mu = \frac{N}{N+1}$, we have

$$\int f^2 \, \mathrm{d}\mu = \frac{N}{(N+1)(\varepsilon - N)}$$

and

$$\int (\Gamma(f) - f^2) d\mu = 1 + \frac{\varepsilon}{(N+1)(N-\varepsilon)}.$$

5.2. Stein's method for generalized Cauchy distributions

In this Section, we shall establish a version of Stein's lemma for generalized Cauchy distributions in dimension one, following the standard approach of explicitly solving a Poisson equation. The result is the following:

THEOREM 5.5 (Stein's method for Cauchy distributions). — Let N < -1, the W_1 distance between a probability measure ν on \mathbb{R} to the generalized Cauchy distribution μ_N^- satisfies

(5.2)
$$W_1(\nu, \mu_N^-) \leqslant \sup \left\{ \int \left[(1+x^2)g'(x) + Nxg(x) \right] d\nu \right\},$$

where the supremum is running over all absolutely continuous function g on \mathbb{R} such that

$$||g||_{\infty} \le \max\left(\frac{4N+3}{|N|(N+1)}, \frac{9}{2}\left(\frac{N}{N+1}\right)^2 + \frac{N}{N+1}\right),$$

and

$$||g'||_{\infty} \le 1 + \left(\frac{3}{2} + \frac{N}{N+1}\right) \frac{N}{N+1}.$$

The constants appearing here are not sharp, but have the sharp order of magnitude as N goes to $-\infty$.

As a corollary, we have the analogue of [35, Theorem 1] for generalized Cauchy target distributions:

COROLLARY 5.6. — Let f be an eigenfunction of a diffusion operator $-\mathcal{L}$ with eigenvalue λ and invariant probability measure μ , and let ν be the pushforward of μ by f. Then for N < -1 we have

$$\begin{split} W_1(\nu,\mu_N^-) \leqslant \left(1 + \left(\frac{3}{2} + \frac{N}{N+1}\right) \frac{N}{N+1}\right) \|\Gamma(f) - f^2 - 1\|_1 \\ + |\lambda + N| \max\left(\frac{4|N| + 3}{|N| \, |N+1|}, \, \frac{9}{2} \left(\frac{N}{N+1}\right)^2 + \frac{N}{N+1}\right) \|f^2\|_1. \end{split}$$

Proof of Corollary 5.6. — Once again, let $\nu = \mu \circ f^{-1}$ and g be a smooth test function.

$$\lambda \int xg(x) d\nu = \lambda \int fg \circ f d\mu = \int (-\mathcal{L}f)g \circ f d\mu$$
$$= \int \Gamma(f, g \circ f) d\mu = \int g' \circ f \Gamma(f) d\mu$$
$$= \int g' \circ f \left(\Gamma(f) - f^2 - 1\right) d\mu + \int (x^2 + 1)g' d\nu.$$

Hence

$$\left| \int \left[(1+x^2)g'(x) + Nxg(x) \right] d\nu \right| \le \|g'\|_{\infty} \|\Gamma(f) - f^2 - 1\|_1 + |\lambda + N|||g||_{\infty} \|f\|_2.$$

We can then apply Theorem 5.5 to conclude.

The proof of Theorem 5.5 relies on Stein's method, as briefly presented in Section 3.3. To begin with, we notice that the probability measure $\mu_N^- := \frac{1}{Z^-}(1+x^2)^{\frac{N}{2}-1}$ on \mathbb{R} , with negative dimension N < -1, is characterized by the integration-by-parts formula

$$\int (1+x^2)g'(x) d\mu_N(x) = -\int Nxg(x) d\mu_N(x).$$

The associated Stein equation is then

(5.3)
$$(1+x^2)g' + Nxg = h - \int h \, \mathrm{d}\mu_N^-$$

for some function h. The solution of interest is given by

(5.4)
$$g(x) = (1+x^2)^{-\frac{N}{2}} \int_{-\infty}^{x} (1+t^2)^{\frac{N}{2}-1} \left(h(t) - \int h \, d\mu_N^-\right) dt$$
$$= -(1+x^2)^{-\frac{N}{2}} \int_{x}^{+\infty} (1+t^2)^{\frac{N}{2}-1} \left(h(t) - \int h \, d\mu_N^-\right) dt,$$

Proof of Theorem 5.5. — For all absolutely continuous h such that $||h||_{Lip} \leq 1$, let g_h be the solution (5.4) of

$$(1+x^2)g'_h + Nxg_h = h - \int h \,\mathrm{d}\mu_N^-.$$

Then we have

$$W_1(\nu, \mu_N^-) = \sup_{\|h\|_{Lip} \leqslant 1} \int \left(h - \int h \, \mathrm{d}\mu_N^- \right) \, \mathrm{d}\nu$$
$$= \sup_{g_h} \int \left((1 + x^2) g_h' + N x g_h \right) \, \mathrm{d}\nu,$$

and bounds on g_h and g'_h given by Lemma 5.8 below complete the proof. \square To prove Lemma 5.8, we need the technical lemma below.

LEMMA 5.7. — We have the following estimates on the cumulative distribution function of μ_N :

$$\forall x < 0, \quad \int_{-\infty}^{x} \frac{1}{Z^{-}} (1 + x^{2})^{\frac{N}{2} - 1} dt \leqslant \min\left(\frac{1}{2}, \frac{1}{|NZ^{-}x|}\right) (1 + x^{2})^{\frac{N}{2}},$$

$$\forall x > 0, \quad \int_{x}^{+\infty} \frac{1}{Z^{-}} (1 + t^{2})^{\frac{N}{2} - 1} dt \leqslant \min\left(\frac{1}{2}, \frac{1}{|NZ^{-}x|}\right) (1 + x^{2})^{\frac{N}{2}}.$$

Moreover.

$$\forall x \le 0, \quad \int_{-\infty}^{x} \frac{1}{C_N} (1+t^2)^{\frac{N}{2}} dt \le \frac{1}{2} (1+x^2)^{\frac{N}{2}+1},$$

where $C_N := \int_{\mathbb{R}} (1+t^2)^{\frac{N}{2}} dt$, and

$$\forall x < 0, \quad \int_{-\infty}^{x} \frac{1}{C_N} (1 + t^2)^{\frac{N}{2}} dt \le \frac{1}{(N+1)C_N x} (1 + x^2)^{\frac{N}{2} + 1}.$$

Proof. — These bounds can be straightforwardly established by studying the functions and their monotonicity.

Let us show the first estimate. Let

$$h(x) := \int_{-\infty}^{x} \frac{1}{Z^{-}} (1 + x^{2})^{\frac{N}{2} - 1} dt - \frac{1}{2} (1 + x^{2})^{\frac{N}{2}}.$$

We have

$$h'(x) = (1+x^2)^{\frac{N}{2}-1} \left[\frac{1}{Z^-} - \frac{N}{2} x \right].$$

Hence h' is decreasing on $(-\infty, \frac{2}{NZ^-})$ and increasing on $(\frac{2}{NZ^-}, 0)$. Now, $h(-\infty) = 0$ and h(0) = 0 because μ_N^- is symetric. Therefore, $h(x) \le 0$. Let $h_1(x) := \int_{-\infty}^x \frac{1}{Z^-} (1+t^2)^{\frac{N}{2}-1} dt - \frac{1}{NZ^{-x}} (1+x^2)^{\frac{N}{2}}$.

Let
$$h_1(x) := \int_{-\infty}^x \frac{1}{Z^-} (1+t^2)^{\frac{N}{2}-1} dt - \frac{1}{NZ^-x} (1+x^2)^{\frac{N}{2}}$$

We have

$$h'_1(x) = \frac{1}{NZ^-x^2}(1+x^2)^{\frac{N}{2}} < 0.$$

Hence $h_1(x) \leqslant h_1(-\infty) = 0$.

The second estimate follows from the first one by symmetry.

With regards to the third estimate, let $h(x) := \int_{-\infty}^{x} \frac{1}{C_N} (1+t^2)^{\frac{N}{2}} dt$ $\frac{1}{2}(1+x^2)^{\frac{N}{2}+1}$. Then

$$h'(x) = (1+x^2)^{\frac{N}{2}} \left(\frac{1}{C_N} - \left(\frac{N}{2} + 1\right)x\right),$$

and there are two cases to consider. Either $N\leqslant -2$ and then h is decreasing on $(-\infty, \frac{N/2+1}{C_N})$, increasing on $(\frac{N/2+1}{C_N}, 0)$ and $h(-\infty) = h(0) = 0$, or $N \in (-2, -1)$ and then h is increasing on \mathbb{R}_- and h(0) = 0. The result therefore stands in both cases.

Finally, let $h_1(x) := \int_{-\infty}^x \frac{1}{C_N} (1+t^2)^{\frac{N}{2}} dt - \frac{1}{(N+1)C_N x} (1+x^2)^{\frac{N}{2}+1}$. Then

$$h_1'(x) = \frac{1}{(N+1)C_N x^2} (1+x^2)^{\frac{N}{2}} < 0,$$

hence h_1 is decreasing, and $h_1(-\infty) = 0$. The final estimate immediately follows.

We shall now establish a priori bounds on solutions to the Poisson equation (5.3).

LEMMA 5.8. — If h is absolutely continuous and μ_N^- -centered, then the solution (5.4) satisfies

$$||g||_{\infty} \leqslant L_N ||h'||_{\infty}$$
 and $||g'||_{\infty} \leqslant K_N ||h'||_{\infty}$

where

$$L_N = \max\left(\frac{4|N|+3}{N(N+1)}, \frac{9}{2}\left(\frac{N}{N+1}\right)^2 + \frac{N}{N+1}\right),$$

and

$$K_N = 1 + \left(\frac{3}{2} + \frac{N}{N+1}\right) \frac{N}{N+1}.$$

Proof. — Let $q(x) := \int_{-\infty}^{x} \frac{1}{Z^{-}} (1+x^2)^{\frac{N}{2}-1} dt$, and let us rewrite the solution g and its derivative. We have

$$h(x) - \int h d\mu_N^- = \int_{-\infty}^x h'(t)q(t) dt - \int_x^{+\infty} h'(t)(1 - q(t)) dt.$$

Combined with (5.4), we get

(5.5)
$$g(x) = -Z^{-} \frac{1 - q(x)}{(1 + x^{2})^{\frac{N}{2}}} \int_{-\infty}^{x} h'(t)q(t) dt - Z^{-} \frac{q(x)}{(1 + x^{2})^{\frac{N}{2}}} \int_{x}^{+\infty} h'(t)(1 - q(t)) dt.$$

Finally, since g is solution to (5.3), we obtain

$$g'(x) = \frac{1}{1+x^2} \left(1 + NZ^- (1 - q(x))x(1+x^2)^{-\frac{N}{2}} \right) \int_{-\infty}^x h'(t)q(t) dt$$
$$-\frac{1}{1+x^2} \left(1 - NZ^- q(x)x(1+x^2)^{-\frac{N}{2}} \right) \int_x^{+\infty} h'(t)(1 - q(t)) dt.$$

Hence

$$||g||_{\infty} \leqslant Z^{-} \sup_{x \in \mathbb{R}} (b_1(x) + b_2(x)) ||h'||_{\infty}$$

and

$$||g'||_{\infty} \le \sup_{x \in \mathbb{R}} (a_1(x) + a_2(x)) ||h'||_{\infty},$$

where

$$a_1(x) := \frac{1}{1+x^2} \left| 1 + NZ^- (1-q(x))x(1+x^2)^{-\frac{N}{2}} \right| \int_{-\infty}^x q(t) dt,$$

$$a_2(x) := \frac{1}{1+x^2} \left| 1 - NZ^- q(x)x(1+x^2)^{-\frac{N}{2}} \right| \int_x^{+\infty} (1-q(t)) dt,$$

$$b_1(x) := (1-q(x))(1+x^2)^{-\frac{N}{2}} \int_{-\infty}^x q(t) dt,$$

and

$$b_2(x) := q(x)(1+x^2)^{-\frac{N}{2}} \int_{-\infty}^x (1-q(t)) dt.$$

It remains to show that

$$K_N := \sup_{x \in \mathbb{R}} (a_1(x) + a_2(x))$$
 and $L_N := Z^- \sup_{x \in \mathbb{R}} (b_1(x) + b_2(x))$

are finite. Let us begin with K_N . Since μ_N is symmetric, $a_1(-x) = a_2(x)$, so $a_1 + a_2$ is symmetric. Moreover, with Lemma 5.7 and since $Z^- \leq C_N$: for all $x \leq 0$,

$$a_{1}(x) = (1+x^{2})^{-1} \int_{-\infty}^{x} q(t) dt + NZ^{-}(1-q(x))x(1+x^{2})^{-\frac{N}{2}-1} \int_{-\infty}^{x} q(t) dt$$

$$\leq (1+x^{2})^{-1} \int_{-\infty}^{x} \frac{1}{2} (1+t^{2})^{\frac{N}{2}} dt$$

$$+ NZ^{-}x(1+x^{2})^{-\frac{N}{2}-1} \int_{-\infty}^{x} \frac{1}{2} (1+t^{2})^{\frac{N}{2}} dt$$

$$\leq \frac{C_{N}}{4} (1+x^{2})^{-1} (1+x^{2})^{\frac{N}{2}+1} + \frac{1}{2} \frac{NZ^{-}}{N+1}$$

$$\leq \frac{C_{N}}{4} + \frac{1}{2} \frac{NZ^{-}}{N+1} \leq \frac{1}{2} C_{N} \left(\frac{1}{2} + \frac{N}{N+1}\right),$$

and

$$\begin{split} \forall x > 0, \quad \int_{-\infty}^{x} q(t) \, \mathrm{d}t &= \int_{-\infty}^{0} q(t) \, \mathrm{d}t + \int_{0}^{x} q(t) \, \mathrm{d}t \\ &\leqslant \frac{1}{2} \int_{-\infty}^{0} (1 + t^{2})^{\frac{N}{2}} \, \mathrm{d}t + \int_{0}^{x} \, \mathrm{d}t \\ &= \frac{C_{N}}{4} + x. \end{split}$$

Hence, for all x > 0,

$$a_1(x) = \left| (1+x^2)^{-1} \int_{-\infty}^x q(t) \, dt + NZ^- (1-q(x))x(1+x^2)^{-\frac{N}{2}-1} \int_{-\infty}^x q(t) \, dt \right|$$

$$\leq (1+x^2)^{-1} \left(\frac{C_N}{4} + x \right) + (1+x^2)^{-1} \left(\frac{C_N}{4} + x \right)$$

$$= 1 + \frac{C_N}{2}.$$

Finally,

$$K_N = \sup_{x \in \mathbb{R}} (a_1(-x) + a_1(x))$$

$$\leq 1 + \frac{1}{2} \left(\frac{3}{2} + \frac{N}{N+1} \right) \int_{\mathbb{R}} (1 + t^2)^{\frac{N}{2}} dt$$

$$\leq 1 + \left(\frac{3}{2} + \frac{N}{N+1} \right) \frac{N}{N+1}.$$

To bound L_N the work is very similar. First, we notice the symmetry $b_1(-x) = b_2(x)$, then using many times again Lemma 5.7, we get

$$\forall x \leqslant -1, \qquad b_1(x) \leqslant \frac{1}{|N|Z^-},$$

$$\forall x \in [-1,0), \qquad b_1(x) \leqslant C_N,$$

$$\forall x \in [0,1], \qquad b_1(x) \leqslant \frac{1}{2} \left(\frac{C_N}{4} + 1\right),$$

$$\forall x \geqslant 1, \qquad b_1(x) \leqslant \frac{-1}{NZ^-} \left(\frac{C_N}{4} + 1\right).$$

So finally

$$\begin{split} L_N &= Z^- \sup_{x \in \mathbb{R}} \left(b_1(x) + b_2(x) \right) \\ &\leqslant \max \left(\frac{1}{|N|} \left(\frac{C_N}{4} + 2 \right) \,, \, Z^- \, C_N + \frac{Z^-}{2} \left(\frac{C_N}{4} + 1 \right) \right), \end{split}$$

which boils down to the result since $Z^- \leqslant C_N \leqslant \frac{2N}{N+1}$.

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