

# Log-Sobolev Inequalities on Metric Spaces

便笺标题

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## 1. Introduction:

- Some background on geometry of metric spaces:

M. Gromov ect.: metric spaces with negative sectional curvature

Bakry - Emery curvature condition:  
for diffusion processes

G. Perelman's papers on Poincaré

Conjecture: Using some results of geometry on metric space with a measure.

Since then, geometry on metric measure spaces was studied by many authors, for example:

J. Lott, K.T. Sturm, S.I. Ohta,  
F. Otto, C. Villani, Alice Chang.

The key points are to extend all important results on Riemannian manifolds to metric spaces.

- The new definition of Ricci curvature of "Markov chains" on metric spaces:

Recently, T. Ollivier gave a new definition of the Ricci curvature,

which extended all the above results for positive Ricci curvature for metric measure spaces, and can be easy to check.

Suppose the Ricci curvature  $\geq R > 0$ , then he proved that a random walk (Markov chain) on metric spaces has the following properties:

Existence and Uniqueness of invariant measure;

$L^1$  and  $L^2$ -Version of Bonnet-Myers theorem;

Poincaré inequality;

Gaussian Concentration (Lévy-Gromov);

## Modified Log-Sobolev inequality

Gromov-Hausdorff continuity of Ricci Curvatures on a sequence of metric spaces.

In particular, he gave an open problem (i.e. problem M) in the lecture in Toulouse that the Log-Sobolev inequality not a modified Log-Sobolev inequality should be proved.

So, the aim of this talk is to prove a standard Log-Sobolev inequality and to consider its applications to Fractal Geometry.

## 2. Definitions and Notations:

- "Markov chain" on metric spaces:

Let  $(X, d)$  be a Polish metric space, a family of probability measures  $m \equiv (m_x : x \in X)$  on  $X$  is called a "random walk" or "Markov chain" on  $X$  if

(i)  $x \in X \rightarrow m_x$  is measurable;

(ii)  $\exists o \in X$  s.t.  $\int d(o, y) m_x(dy) < \infty, (\forall x \in X)$ .

- Wasserstein transportation distance:

Let  $\nu_1, \nu_2$  be two probability measures on  $X$ ,  $\pi(\nu_1, \nu_2)$  be the set of all coupling measures of  $\nu_1$  and  $\nu_2$ , and then the Wasserstein transportation distance between  $\nu_1$  and  $\nu_2$  is defined by

$$W_1(\nu_1, \nu_2) = \inf_{\nu \in \pi(\nu_1, \nu_2)} \int_{X \times X} d(x, y) \nu(dx dy).$$

- Ricci curvature of  $m = (m_x : x \in X)$ :  
The Ricci curvature of  $(X, d, m)$  along  $(x, y)$ ,  $x, y \in X$ , is defined by
 
$$k(x, y) = 1 - \frac{\mathcal{J}_1(m_x, m_y)}{d(x, y)} .$$

### Remarks:

- $m = (m_x : x \in X)$  is just a probabilistic transition kernel for a Markov chain, and  $k(x, y)$  is just the coefficients of contraction of the Markov chain in transportation distance, it is widely used by Dobrushin, M.F. Chen, etc., especially appeared in "Dobrushin criterion" for spin systems.
- In special cases, some results about spectral gap (e.g. M.F. Chen, F.T. Wang), Concentration of measures (e.g. K. Marton,

Djellout — Guillin — L.M. Wu) were proved.

• There are a lot of examples about random walk  $m$  with positive curvature, for example:

①. Metric measure space  $(X, d, \mu)$  **With positive Ricci curvature**  
 $\text{supp}(\mu) = X,$

$$m_x = \mu|_{B(x, \varepsilon)} / \mu(B(x, \varepsilon)), \quad \varepsilon > 0.$$

In particular,  $X$  is a **Riemannian manifold with positive Ricci curvature.**

②. Simple random walk on  $\mathbb{Z}^N$ , where  $k(x, y) = 0$ .

③. Discrete 0-1 process on

$$X = \{-N, \dots, N\}, \text{ where}$$

$$x \sim y \Rightarrow k(x, y) = \frac{1}{2N}.$$

④. **Diffusion process  $(X_t)$  on Riemannian manifold** with

$$dX_t = F dt + dB_t,$$

where, Bakry-Emery curvature for  $L = \frac{1}{2}\Delta + F \cdot \nabla$  is  $\frac{1}{2}\text{Ric}_M - \nabla^{\text{sym}} F$ , and when  $d(x, y)$  is small enough, then  $k(x, y)$  is determined by  $\frac{1}{2}\text{Ric}_M - \nabla^{\text{sym}} F$ .

⑤. Kac's random walk on  $SO(N)$ , where  $k(x, y) \sim \frac{1}{N^2}$ .

⑥. Glauber Dynamics for the Ising model over a finite graph.

⑦.  $M/M/\infty$  queues.

Notation  $S$ :

Let  $(X, d)$  be a Polish metric space, and  $m = (m_x : x \in X)$  be a random walk,

$$\sigma_\infty(x) = \frac{1}{2} \text{diam Supp}(m_x),$$

$$\sigma_\infty = \sup_{x \in X} \sigma_\infty(x);$$



•  $\lambda$ -range gradient of  $f: X \rightarrow \mathbb{R}: (\lambda > 0)$

$$(Df)(x) \triangleq (D_\lambda f)(x) = \sup_{y, y' \in X} \frac{|f(y) - f(y')|}{d(y, y')} e^{-\lambda d(x, y) - \lambda d(x, y')};$$

If  $f$  is smooth and  $X$  is a compact Riemannian manifold, then

$$(D_\lambda f)(x) \rightarrow |Df(x)| \text{ as } \lambda \rightarrow +\infty.$$

• averaging operator:

$$Mf(x) = \int f(y) m_x(dy).$$

• variance and entropy of  $f: X \rightarrow \mathbb{R}$ :

let  $\nu$  be any probability measure on  $X$ , then

$$\text{Var}_\nu(f) = \frac{1}{2} \int |f(x) - f(y)|^2 \nu(dx) \nu(dy),$$

$$\text{Ent}_\nu(f) = \int f \log \frac{f}{\|f\|_{L^1(\nu)}} d\nu, \quad f \geq 0.$$

⊗ Unstability:  $U(x, y) = \frac{R(x, y)}{R(x, y)}$

$$U = \sup_{x \neq y} U(x, y).$$

### 3. Y. Ollivier's Result on LSI:

In his paper, Y. Ollivier proved the following modified LSI:

Let  $m = (m_x : x \in X)$  be a random walk on  $(X, d)$ ,  $\sigma_\infty < +\infty$  and  $k(x, y) \geq k > 0$ . Then for small enough

$\lambda > 0$  we have:  $\forall f \geq 0$ ,

$$\text{Ent}_\mu(f) \leq \frac{4}{k} C_m \int \frac{(Df)^2}{f} d\mu,$$

where  $\mu$  is the invariant measure of  $m$ ,

and

$$C_m = \sup \left\{ \text{Var}_{m_x}(f) : \begin{array}{l} f \text{ is } 1\text{-Lipschitz} \\ x \in X \end{array} \right\}.$$

In his proof, he proved or used following important results:

A. Associativity of entropy :

$$\bar{\text{Ent}}_{\mu} f = \sum_{t \in \mathbb{Z}_+} \int_X \text{Ent}_{m_x} (M^t f) \mu(dx),$$

( e.g. see Th. D.13 in Dembo - Zeitouni 's book on Large deviations );

$$\text{Ent}_{m_x} f \leq \frac{1}{Mf(x)} \text{Var}_{m_x} f,$$

( using  $a \log a \leq a^2 - a$  ).

B. Gradient contraction :

for  $\lambda$  small enough,  $Df < +\infty$ ,

$$(DMf)(x) \leq (1 - \frac{\lambda}{2}) M(Df)(x).$$

C. If  $Df < +\infty$ , then  $\forall x \in X$

$$|f(y) - f(z)| \leq d(y, z) e^{4\lambda \sigma_{\infty} M(Df)(x)},$$

$\forall y, z \in \text{Supp}(m_x)$ .

#### 4. Standard LSI :

using the facts A, B, c in the above and the following Lemma that we proved

Lemma D: Let  $D_{\lambda} f < +\infty$ , then  
 $\forall x \in X$  we have :

$$D_{\lambda} f^2(x) \leq 2 f(x) D_{\lambda} f(x) + \frac{2}{e\lambda} (D_{\lambda} f)^2(x);$$

we proved the following standard Log-Sobolev inequality :

**Theorem A:** Let  $m = (m_x : x \in X)$

be a random walk on  $(X, d)$ ,

$\sigma_{\infty} < +\infty$ ,  $k(x, y) \geq k > 0$  ( $\forall x, y \in X$ )

Then for small enough  $\lambda > 0$ ,

there is a universal constant  $N > 0$

such that:  $\forall f \in L^2(\mu)$ ,

$$\text{Ent}_\mu(f^2) \leq \frac{N \sigma_\infty^2(HU)}{k} \int (Df)^2 d\mu$$

Moreover, for any  $\varepsilon \in [1, 2]$  there is a positive constant  $C(\varepsilon, k) > 0$  such that:

$\forall f \in L^1(\mu)$  and  $f \geq 0$   $\mu$ -a.e.,

$$\text{Ent}_\mu(f) \leq C(\varepsilon, k) \int \left| \frac{Df}{f} \right|^\varepsilon f d\mu.$$

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Remark: In fact, we can prove

that:

$$N \leq 2^5 \times 3^2 \times 5^2 \times e^{-\frac{14}{15}}$$

$$C(\varepsilon, k) \leq \frac{\sqrt{2}}{k} \sup_{x \in X} \left( \int_{X \times X} d(y, z)^{\frac{\varepsilon}{\varepsilon-1}} m_x(dy) m_x(dz) \right)^{\varepsilon-1}$$

The above results are the improving for corresponding results in J. Ollivier's paper.

We also improve Y. Ollivier's Gaussian concentration result by avoiding the condition  $\sigma_\infty < +\infty$ :

Proposition: Let  $\varphi: X \rightarrow \mathbb{R}$  be a

$\alpha$ -Lipschitz function with  $\alpha \leq 1$ .

Assume  $\sup_x \frac{\sigma(x)^2}{n_x} \leq B^2$  and  $0 \leq \lambda \leq \frac{1}{2B}$ .

Then for  $x \in X$

$$(M e^{\lambda \varphi})(x) \leq e^{\lambda M \varphi(x) + 4 \lambda^2 \alpha^2 \frac{\sigma(x)^2}{n_x}}$$

Moreover, if  $x \mapsto \frac{\sigma(x)^2}{n_x k}$  is  $C$ -Lipschitz,

then for  $0 \leq \lambda \leq \frac{1}{2B} \wedge \frac{1}{8C}$ ,

$$\mathbb{E}_\mu e^{\lambda(f - \mathbb{E}_\mu f)} \leq e^{\frac{16}{3} \lambda^2 \mathbb{E}_\mu \frac{\sigma(x)^2}{n_x k}}$$

for any 1-Lipschitz function  $f$ . #

## 5. Applications for Fractal Geometry:

⊠. Let  $\mu$  be a Borel probability measure on  $(X, d)$ , and  $(\mathcal{E}, \mathcal{F})$  be a regular Dirichlet form in  $L^2(\mu)$ . Define

$$R(x, y) = \sup_{\mathcal{E}(f, f) \neq 0} \frac{|f(x) - f(y)|^2}{\mathcal{E}(f, f)}, \quad x, y \in X.$$

$R(\cdot, \cdot)$  is so called "effective resistance" of  $(\mathcal{E}, \mathcal{F})$ .

We assume that:

- $R(x, y) < +\infty, \forall x, y \in X$ ;
- $R(\cdot, \cdot)^{\frac{1}{2}}$  is a distance on  $X$ , and it induces the same topology as  $d$ .

In this case, we called  $(\mathcal{E}, \mathcal{F})$ 's a resistance form. In fractal geometry there are many Dirichlet forms (e.g. Dirichlet form associated with Brownian

motions on many fractal sets) are resistance forms.

By the results in [Gong - Wu 06] we know that, spectral gap property of  $(\mathcal{E}, \mathcal{F})$  usually concerns with the resolvents of  $(\mathcal{E}, \mathcal{F})$ . Hence, we define the following measure kernels

$m = (m_x : x \in X)$  on  $(X, \mathcal{R}^{\frac{1}{2}})$ : For  $\alpha > 0$ ,  
 $m_x = \alpha G_\alpha(x, \cdot) = \alpha \int_0^\infty e^{-\alpha t} p_t(x, \cdot) dt, \forall x \in X.$

Then  $m = (m_x : x \in X)$  is a random walk on  $(X, \mathcal{R}^{\frac{1}{2}})$  if  $(\mathcal{E}, \mathcal{F})$  is conservative.

We can easily prove that

**Theorem B:**  $(X, \mathcal{R}^{\frac{1}{2}}, m)$  has **positive Ricci curvature** if  $1 - 2\alpha \int R(0, x) \mu(dx) < 1$ .

1). If  $\sup_{x, y} R^{\frac{1}{2}}(x, y) = D < +\infty,$



then the Markov process  $(X_t)$  associated with  $(\mathcal{E}, \mathcal{F})$  has a unique invariant measure  $\mu$ , and the following log-Sobolev inequality holds:  $\forall f \in \mathcal{F}$

$$\text{Ent}_\mu(f^2) \leq 2D^2 \mathcal{E}(f, f).$$

2). If  $\int R(0, x) \mu(dx) < +\infty$ , then the Poincaré inequality holds:  $\forall f \in \mathcal{F}$

$$\text{Var}_\mu(f) \leq C_p \mathcal{E}(f, f)$$

with  $C_p \leq \int R(0, x) \mu(dx)$ . Moreover,

we have the following Gaussian concentration: for  $f \in \mathcal{F}$  with  $\mathcal{E}(f, f) \leq 1$

$$\mathbb{E}_\mu e^{\lambda(f - \mathbb{E}_\mu f)} \leq e^{4C_p \lambda^2}$$

for any  $0 \leq \lambda \leq \frac{1}{2\sqrt{C_p}}$ .  $\#$

△ For Theorem B 2) we have the following example:  $X = \mathbb{R}$ ,  $\mu(dx) = e^{-cx^\alpha} dx$  for  $\alpha > 2$ ,  $\mathcal{E}(f, f) = \int |f'|^2 d\mu$ . Then

$$\int R(0, x) \mu(dx) < +\infty, \quad \sup_{x, y} R(x, y) = +\infty.$$

△ For Theorem B 1) we have the following important examples in fractal geometry:

△ Let  $(X, d)$  be the **Sierpinski gasket**,  $\mu$  be the normalized Hausdorff measure on  $X$ , and  $(\mathcal{E}, \mathcal{F})$  be given by Example 3.1.5 in J. Kigami's book "Analysis on fractals" (Cambridge, 2001), then Theorem B1) holds with  $D^2 \leq \frac{6}{5}$ .

Similar results hold for other post-critical-finite sets with regular harmonic structures.

⊗). Let  $(X, d)$  be the **Sierpinski carpet**,  $\mu$  be the normalized Hausdorff measure on  $X$ , and  $(\mathcal{E}, \mathcal{F})$  be the self-similar Dirichlet form defined by S. Kusuoka and X.Y. Zhou in 1992, PTRF, then Theorem B1) holds.

⊗). Let  $X$  be a self-similar set in a locally compact Polish space  $(K, d_0)$  with  $d_f < 2$ , which satisfies open set condition and admits a geodesic metric  $d_r \sim d_0^\theta$  for some  $\theta > 0$ , and  $\mu$  be the normalized Hausdorff measure on  $X$ . Then by the results in K.T. Sturm's paper on Ann. Prob. 1998 we can construct a Dirichlet form  $(\mathcal{E}, \mathcal{F})$  in  $L^2(\mu)$ . For this  $(\mathcal{E}, \mathcal{F})$ , Theorem B1) also holds. #

Note: The abstract of Y. Ollivier's paper has published on C. R. Math. Acad. Sci. Paris 345 (2007) n° 11.

Thank you!

