

*Measure concentration, functional inequalities,  
and curvature of metric measure spaces*

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circle of ideas

between analysis, geometry and probability theory

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**concentration of measure phenomenon**

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**(Ricci) curvature bounds on metric measure spaces**

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$$K \cap F$$

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normalized volume  $\mu(\cdot) = \frac{\text{vol}(\cdot)}{\text{vol}(\mathbb{S}^n)}$

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Gauss space : curvature 1 dimension  $\infty$

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infinite dimensional analysis (Wiener space)

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complementary to PDE and calculus of variations viewpoint

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$u = \chi_A, \quad v = \chi_B$  multiplicative form of Brunn-Minkowski

$$\text{vol}_n(\theta A + (1 - \theta)B) \geq \text{vol}_n(A)^\theta \text{vol}_n(B)^{1-\theta}$$

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extension :  $d\mu = e^{-V} dx$ ,  $V'' \geq c > 0$

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$$f \geq 0 \text{ smooth on } \mathbb{R}^n, \int f d\gamma = 1$$

$$\int f \log f d\gamma \leq \frac{1}{2} \int \frac{|\nabla f|^2}{f} d\gamma$$

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A. Stam (1959), L. Gross (1975)

concentration via the logarithmic Sobolev inequality (I. Herbst)

$$F : \mathbb{R}^n \rightarrow \mathbb{R} \quad \text{1-Lipschitz,} \quad \int F d\gamma = 0$$

$$f = e^{\lambda F} / \int e^{\lambda F} d\gamma, \quad \lambda \in \mathbb{R}$$

differential inequality on  $\int e^{\lambda F} d\gamma$

$$\int e^{\lambda F} d\gamma \leq e^{\lambda^2/2}, \quad \lambda \in \mathbb{R}$$

$$\gamma(\{F < r\}) \geq 1 - e^{-r^2/2}, \quad r > 0$$

**measure/information theoretic description :**

transportation cost inequality

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$$W_2(\nu, \gamma)^2 = \inf_{\nu \leftarrow \pi \rightarrow \gamma} \int \int \frac{1}{2} |x - y|^2 d\pi(x, y)$$

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**M. Talagrand (1996)**

concentration via the transportation cost inequality (K. Marton)

$$A, B \subset \mathbb{R}^n, \quad d(A, B) \geq r > 0$$

$$\gamma_A = \gamma(\cdot | A), \quad \gamma_B = \gamma(\cdot | B)$$

$$W_2(\gamma_A, \gamma_B) \leq \left( \log \frac{1}{\gamma(A)} \right)^{1/2} + \left( \log \frac{1}{\gamma(B)} \right)^{1/2}$$

$$\frac{r}{\sqrt{2}} \leq W_2(\gamma_A, \gamma_B) \leq \left( \log \frac{1}{\gamma(A)} \right)^{1/2} + \left( \log \frac{1}{\gamma(B)} \right)^{1/2}$$

$$\gamma(A) \geq 1/2, \quad B = \text{complement of } A_r$$

$$\gamma(A_r) \geq 1 - e^{-r^2/4}, \quad r \geq r_0$$

Prékopa-Leindler inequality

logarithmic Sobolev inequality

transportation cost inequality

Prékopa-Leindler inequality

$$\int w d\gamma \geq \left( \int u d\gamma \right)^\theta \left( \int v d\gamma \right)^{1-\theta}$$

logarithmic Sobolev inequality

$$\int f \log f d\gamma \leq \frac{1}{2} \int \frac{|\nabla f|^2}{f} d\gamma$$

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$$\int w d\mu \geq \left( \int u d\mu \right)^\theta \left( \int v d\mu \right)^{1-\theta}$$

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$$\int f \log f d\mu \leq \frac{C}{2} \int \frac{|\nabla f|^2}{f} d\mu$$

transportation cost inequality

$$W_2(\nu, \mu)^2 \leq C H(\nu | \mu)$$

$\mu$  probability measure on  $\mathbb{R}^n$

or more general spaces

# hierarchy

Prékopa-Leindler inequality



logarithmic Sobolev inequality



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**F. Otto, C. Villani (2000)**

PDE and transportation methods

hypercontractivity of Hamilton-Jacobi equations



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**dimension free measure concentration**

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tools to establish these inequalities

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**parametrisation methods**

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equivalent to a curvature condition

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## extensions

- ◇ second order differential operators

**D. Bakry, M. Emery (1985)**

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**Y. Ollivier (2008)**



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manifold case R. McCann (1995)



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non-smooth analysis, PDE methods

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extends to (weighted) manifolds

characterizes  $\text{Ric} + \text{Hess}(V) \geq c$

**K. Th. Sturm (2005)**

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in a metric measure space (length space)  $(X, d, \mu)$

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definition of lower bound on curvature

**postulate** that entropy is  $c$ -convex along one geodesic  $(\mu_\theta)_{\theta \in [0,1]}$

## notion of Ricci curvature bound

in a metric measure space (length space)  $(X, d, \mu)$

$(\mu_\theta)_{\theta \in [0,1]}$  geodesic in  $(\mathcal{P}_2(X), W_2)$  connecting  $\mu_0, \mu_1$

definition of lower bound on curvature

**postulate** that entropy is  $c$ -convex along one geodesic  $(\mu_\theta)_{\theta \in [0,1]}$

$$H(\mu_\theta | \mu) \leq (1 - \theta)H(\mu_0 | \mu) + \theta H(\mu_1 | \mu) - c \theta(1 - \theta)W_2(\mu_0, \mu_1)^2$$

$H$  relative entropy,  $W_2$  Wasserstein distance

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in metric measure space

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- ◇ generalizes Ricci curvature in Riemannian manifolds

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- ◇ allows for geometric and functional inequalities  
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  - ◇ allows for geometric and functional inequalities (transportation cost, logarithmic Sobolev inequalities)
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