

Therefore,

$$R^\sigma(h) = \frac{\sum_{\{u,v\}} (h_{uv} - \sigma_{uv} h_{uv})^2}{\sum_u h_{uv}^2 du} \leq \frac{2 \sum_{i=1}^k \sum_{\{u,v\}} (c_i f_i(u) - \sigma_{uv} c_i f_i(v))^2}{\sum_u \sum_{i=1}^k c_i^2 f_i^2(u) du}$$

$$\leq 2 \max_{1 \leq i \leq k} R^\sigma(f_i).$$

□.

Remark: Observing  $\sqrt{\frac{2}{\lambda_k^\sigma}} \geq 1$ , Theorem II.9.1 follows directly from Theorem  $\textcircled{*}$ .

We also have a higher order version of Thm II.9.1 and Thm II.9.6  $\textcircled{*}$ .

Theorem II.9.8 There exists an absolute constant  $C$  such that for any signed graph  $\Gamma = (G, \sigma)$  and any  $1 \leq k \leq l \leq |V|$ , at least one of the two estimates holds:

(i)  $h_k^\sigma \leq 8C l k^6 \lambda_k^\sigma$ .

(ii)  $h_k^\sigma < 16\sqrt{2} C l k^6 \frac{\lambda_k^\sigma}{\sqrt{\lambda_l^\sigma}}$ .

As a consequence, for any  $\Gamma = (G, \sigma)$  and any  $1 \leq k \leq l \leq |V|$ , we have

$$h_k^\sigma < 16\sqrt{2} C l k^6 \frac{\lambda_k^\sigma}{\sqrt{\lambda_l^\sigma}}$$

Proof: We need the following fact from random partition theory: For any  $k \in \{1, 2, \dots, |V|\}$ , there exists  $k$  disjointly supported functions  $\psi_1, \psi_2, \dots, \psi_k : V \rightarrow \mathbb{R}$  s.t.

$$R^\sigma(\psi_i) \leq C k^6 R^\sigma(\Phi) \leq C k^6 \lambda_k^\sigma, \quad i=1, 2, \dots, k, \quad \textcircled{**}$$

where  $C$  is an absolute constant. Recall  $\Phi$  is the map (146)  
 from  $V$  to  $\mathbb{R}^k$  given by the first  $k$  eigenfunctions.

Next, we apply Lemma II.9.5 to each  $\psi_i$ : For each  $\psi_i$ , for any  $k \leq \ell \leq |V|$ , at least one of the following 2

holds:

(i)  ~~$\phi^\sigma(\psi_i)$~~   $\phi^\sigma(\psi_i) \leq 8\ell R^\sigma(\psi_i)$

(ii)  $\frac{\lambda_\ell^\sigma}{2} < 256\ell^2 \frac{R^\sigma(\psi_i)^2}{\phi^\sigma(\psi_i)^2}$ .

Note that, we already apply Lemma II.9.7 in (ii) for the ease of notation. Since  $\psi_i$ 's are disjointly supported, we observe that

$$h_k^\sigma \leq \phi^\sigma(\psi_i), \quad \forall i=1,2,\dots,k.$$

We then further apply the fact  $(*)$  and derive at least one of the two holds:

(i)  $h_k^\sigma \leq 8\ell \cdot Ck^6 \lambda_k^\sigma$       (ii)  $\frac{\lambda_\ell^\sigma}{2} < 256\ell^2 \frac{(Ck^6 \lambda_k^\sigma)^2}{(h_k^\sigma)^2}$ .

That is, at least one of the following two holds:

(i)  $h_k^\sigma \leq 8C\ell k^6 \lambda_k^\sigma$ ,      (ii)  $h_k^\sigma < 16\sqrt{2} C\ell k^6 \frac{\lambda_k^\sigma}{\sqrt{\lambda_\ell^\sigma}}$ .

Using the fact  $\lambda_\ell^\sigma \leq 2$ , (ii) implies

$$h_k^\sigma \leq 8\sqrt{2} C\ell k^6 \frac{\lambda_k^\sigma}{\sqrt{\lambda_\ell^\sigma}}. \quad \square$$

## (II.10) Buser inequality

(147)

In last section, we discuss when the lower bound estimates of Cheeger inequalities ~~to~~ have the same order as its upper bound estimates. In this section, we consider a "dual" question: When does the upper bound estimates have the same order as its lower bounds? That is, under which conditions, <sup>can</sup> it hold that

$$\lambda_k^\sigma \lesssim (h_k^\sigma)^2?$$

One ~~such~~ answer in this direction is Buser type inequalities, imposing certain curvature ~~conditions~~ restrictions of the underlying graph. Our aim is to show the following:

Theorem II.10.1: For any signed graph  $\Gamma = (G, \sigma)$  satisfying  $CD^\sigma(0, \infty)$  condition, and any  $k \in \{1, 2, \dots, |V|\}$ , we have

$$\lambda_k^\sigma \leq 16k^2 \log(2k) d_{\max} (h_k^\sigma)^2.$$

(L., Münch, Peyerimhoff, SIAM J. Discrete Math. 33 (2019), no. 1257-305)

### Curvature of a signed graph.

The motivation for the curvature-type condition  $CD^\sigma(0, \infty)$  ~~is~~ lies in the Bochner identity in Riemannian geometry. Recall on a Riemannian manifold  $(M^n, g)$ , for any  $f \in C^\infty(M)$ , we have

$$\frac{1}{2} \Delta |\nabla f|^2 = |\text{Hess} f|^2 + \langle \nabla \Delta f, \nabla f \rangle + \text{Ric}(\nabla f, \nabla f).$$

The Ricci curvature term  $\text{Ric}(\nabla f, \nabla f)$  appears when one 148 try to commute the order of <sup>taking</sup> covariant derivatives using Ricci identity. By Cauchy-Schwarz inequality, we have

$$|\text{Hess } f|^2 \geq \frac{1}{n} (\Delta f)^2.$$

Indeed, we have the following characteristic of Ricci curvature lower bound:

On  $(M^n, g)$ , let  $x \in M$ , then  $\text{Ric} \geq K$  at  $x$  iff

$$\textcircled{*} \quad \frac{1}{2} \Delta |\nabla f|^2(x) \geq \frac{1}{n} (\Delta f)_{(x)}^2 + \langle \nabla \Delta f, \nabla f \rangle_{(x)} + K |\nabla f|^2_{(x)}, \text{ holds at } x$$

for all  $f \in C^\infty(M)$ .

Then a key observation due to Bakry and Émery in 1980s is as follows: One can rewrite  $\textcircled{*}$  as

$$\frac{1}{2} \Delta |\nabla f|^2(x) - \langle \nabla \Delta f, \nabla f \rangle_{(x)} \geq \frac{1}{n} (\Delta f)_{(x)}^2 + K |\nabla f|^2(x). \quad (1)$$

Compare it with the classical chain rule formula:

$$\frac{1}{2} \Delta (f^2)_{(x)} - f \Delta f_{(x)} = |\nabla f|^2(x). \quad (2)$$

The LHSs of (1) and (2) share the same pattern.

Definition: <sup>II.10.2</sup> (Bakry - Émery) For any  $f, g \in C^\infty(M)$ , we define

$$\Gamma(f, g)(x) := \frac{1}{2} (\Delta(fg)(x) - f(x)\Delta g(x) - \Delta f(x) \cdot g(x))$$

$$\text{and } \Gamma_2(f, g)(x) := \frac{1}{2} (\Delta \Gamma(f, g)(x) - \Gamma(f, \Delta g)(x) - \Gamma(\Delta f, g)(x)).$$

Remark: One may denote  $\Gamma_0(f, g)(x) := f(x)g(x)$ . Then  $\Gamma_2(f, g)(x)$  is obtained by replacing  $\Gamma_0$  by  $\Gamma$  in the definition of  $\Gamma(f, g)(x)$ .

In principle, one can go on & defining  $\Gamma_3, \Gamma_4, \dots$ . (149)

Observe that, LHS of (2) is  $\Gamma(f, f)(x) := \Gamma(f)(x)$ , while the LHS of (1) is  $\Gamma_2(f, f)(x) := \Gamma_2(f)(x)$ .

The characterization of  $\text{Ric} \geq K$  at  $x$  in  $(M^n, g)$  can be reformulated as belows:

⊛  $\Gamma_2(f)(x) \geq \frac{1}{n}(\Delta f)^2 + K \Gamma(f)(x)$ , holds at  $x$  for any  $f \in C^\infty(M)$ .

In this way, one can talk about "curvature" of an operator.

⊛' is also called the curvature-dimension inequality  $\text{CD}(K, n)$ .

Particularly, we can define  $\text{CD}(K, n)$  for Laplacian on a graph.

Definition <sup>II.10.2</sup>. Let  $G = (V, E)$  be a graph.  $\Gamma, \Gamma_2$  are defined as in Definition II.10.2 via the graph Laplacian. Then  ~~$G$  is said~~

We say  $G$  satisfies  $\text{CD}(K, n)$  at  $x \in V$  if

$$\Gamma_2(f)(x) \geq \frac{1}{n}(\Delta f(x))^2 + K \Gamma(f)(x), \text{ for any } f: V \rightarrow \mathbb{R}.$$

Prop II.10.4. For any  $f: V \rightarrow \mathbb{R}$ , we have at  $x \in V$

$$\Gamma(f)(x) = \frac{1}{2} \frac{1}{dx} \sum_{y: y \sim x} (f(y) - f(x))^2$$

Proof: This follows directly by definition:

$$\begin{aligned} \Gamma(f)(x) &= \frac{1}{2} \Delta(f^2)(x) - f(x) \Delta f(x) \\ &= \frac{1}{2} \frac{1}{dx} \sum_{y: y \sim x} (f^2(y) - f^2(x)) - f(x) \cdot \frac{1}{dx} \sum_{y: y \sim x} (f(y) - f(x)) \\ &= \frac{1}{2} \frac{1}{dx} \sum_{y: y \sim x} (f^2(y) - f^2(x) - 2f(x)f(y) + 2f^2(x)) \\ &= \frac{1}{2} \frac{1}{dx} \sum_{y: y \sim x} (f(y) - f(x))^2. \quad \square \end{aligned}$$

Remark: This fits with our previous discussions very well. For example,

$$\begin{aligned} \sum_{\{x,y\}} (f(y) - f(x))^2 &= \frac{1}{2} \sum_{x \in V} \sum_{y: y \sim x} (f(y) - f(x))^2 \\ &= \sum_{x \in V} \frac{1}{2 d_x} \sum_{y: y \sim x} (f(y) - f(x))^2 d_x \\ &= \sum_{x \in V} \Gamma(f)(x) d_x \end{aligned}$$

The Rayleigh quotient

$$R(f) = \frac{\sum_{x \in V} \Gamma(f)(x) d_x}{\sum_{x \in V} f(x)^2 d_x}$$

Next, we look for ~~some~~ <sup>The</sup> formulation of  $\Gamma, \Gamma_2$  which fits neatly with the signed Laplacian! <sup>Most natural idea is to</sup> use  $\Delta^\sigma$  instead of  $\Delta$  in the definitions of  $\Gamma, \Gamma_2$ , and  $CD(K, n)$ . However, this does not work. The proper way to do it is as follows.

Definition II.10.5. (Curvature-dimension ineq. for signed Laplacian)

For any two functions  $f, g: V \rightarrow \mathbb{R}$ , we define at  $x \in V$

$$\Gamma^\sigma(f, g)(x) := \frac{1}{2} \left( \Delta(fg)(x) - f(x)\Delta^\sigma g(x) - \Delta^\sigma f(x) \cdot g(x) \right),$$

$$\text{and } \Gamma_2^\sigma(f, g)(x) := \frac{1}{2} \left( \Delta \Gamma^\sigma(f, g)(x) - \Gamma^\sigma(f, \Delta^\sigma g)(x) - \Gamma^\sigma(\Delta^\sigma f, g)(x) \right)$$

We say ~~a~~ a signed graph  $\mathcal{G} = (G, \sigma)$  satisfies ~~the~~  $CD^\sigma(K, n)$  at  $x \in V$  if

$$\Gamma_2^\sigma(f)(x) \geq \frac{1}{n} (\Delta^\sigma f(x))^2 + K \Gamma^\sigma(f)(x), \text{ holds for any } f: V \rightarrow \mathbb{R}.$$

Prop II.10.6. For any  $f: V \rightarrow \mathbb{R}$ , we have at  $x \in V$

$$\Gamma^\sigma(f)(x) = \frac{1}{2} \frac{1}{d_x} \sum_{y: y \sim x} (\sigma_{xy} f(y) - f(x))^2.$$

Proof: This again follows directly from definition:

(157)

$$\begin{aligned} \Gamma^\sigma(f)(x) &= \frac{1}{2} \Delta(f^2)(x) - f(x) \Delta^\sigma f(x) \\ &= \frac{1}{2} \frac{1}{dx} \sum_{y: y \sim x} (f^2(y) - f^2(x)) - f(x) \cdot \frac{1}{dx} \sum_{y: y \sim x} (f(y) - f(x)) \\ &= \frac{1}{2} \frac{1}{dx} \sum_{y: y \sim x} (f^2(y) - f^2(x) - 2\sigma_{xy} f(x)f(y) + 2f(x)^2) \\ &= \frac{1}{2} \frac{1}{dx} \sum_{y: y \sim x} (\sigma_{xy} f(y) - f(x))^2 \quad \square \end{aligned}$$

Rmk: The corresponding Rayleigh quotient can be reformulated

as: 
$$R^\sigma(f) = \frac{\sum_{x \in V} \Gamma^\sigma(f)(x) dx}{\sum_{x \in V} f(x)^2 dx} \quad \square$$

Prop II.10.7: The  $CD^\sigma$  inequality is switching invariant. Let  $\Gamma = (G, \sigma)$  be a signed graph. Let  $\tau: V \rightarrow \{+1, -1\}$  be a switching function. Then  $(G, \sigma^\tau)$  satisfies  $CD^{\sigma^\tau}(K, n)$  iff  $(G, \sigma)$  satisfies  $CD^\sigma(K, n)$ .

Proof: Observe that <sup>we have</sup> as matrices (Recall  $\sigma_{xy}^\tau = \tau(x)\sigma_{xy}\tau(y)$ ).

$$\begin{aligned} \Delta^{\sigma^\tau} &= \begin{pmatrix} \tau(x_1) & & \\ & \ddots & \\ & & \tau(x_n) \end{pmatrix} \Delta^\sigma \begin{pmatrix} \tau(x_1) & & \\ & \ddots & \\ & & \tau(x_n) \end{pmatrix} \\ &=: M_\tau^{-1} \Delta^\sigma M_\tau. \end{aligned}$$

Furthermore, for any  $f, g: V \rightarrow \mathbb{R}$ , we have

$$\Gamma^{\sigma^\tau}(f, g) = \Gamma^\sigma(\tau f, \tau g), \quad \Gamma_2^{\sigma^\tau}(f, g) = \Gamma_2^\sigma(\tau f, \tau g).$$

Then the proposition follows immediately.  $\square$

Rmk:  $CD^\sigma(K, n)$  at  $x$  is a local property: Only the structure of  $B_2(x)$  is related to ~~the~~  $CD^\sigma(K, n)$  property at  $x$ .

# Heat semigroup characterizations

We discuss a very useful characterization of  $C^0(K, \infty)$  for our later purpose.

Consider the following continuous-time heat equation:

$$\begin{cases} \frac{\partial u(x,t)}{\partial t} = \Delta^\sigma u(x,t) \\ u(x,0) = f(x) \end{cases} \quad (3)$$

where  $f: V \rightarrow \mathbb{R}$  is an initial function.

The equation (3) has a unique solution according to the Picard-Lindelöf theorem. Applying Picard iteration, we know this unique solution ~~can be~~ is given by the matrix exponential.

$$u(x,t) = \left( e^{t\Delta^\sigma} f \right) (x).$$

↳ considered as a  $VT$ -vector.

• A brief recall of matrix exponential:

(a) For any square matrix  $A$ ,

$$e^A := \sum_{n=0}^{\infty} \frac{A^n}{n!}, \text{ where } A^0 = I.$$

The serie is always converge. (entry-wise).

(b)  $\frac{d}{dt} e^{At} = A \cdot e^{At}$

(c)  $e^{\mathbf{0}} = I$  ← zero matrix.

(d)  $e^A \cdot e^B = e^{A+B}$  if  $AB = BA$ .

(e)  $B e^{AB} = e^A \cdot B$  if  $AB = BA$ .

(f)  $(e^A)^T = e^{A^T}$

(g)  $e^A = P^{-1} e^{PAP^{-1}} P$ , for  $P$  invertible.



For diagonal matrices, the matrix exponential has explicit expression: (153)

$$e^{\begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix}} = \begin{pmatrix} e^{\lambda_1} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n} \end{pmatrix}. \quad \square$$

Prop II.10.8: The operator  $P_t^\sigma := e^{t\Delta^\sigma}$ ,  $t \geq 0$ :  $\ell^2(V) \rightarrow \ell^2(V)$

satisfies the following properties:

(i)  $P_t^\sigma$  is self-adjoint.

(ii)  $P_t^\sigma \Delta^\sigma = \Delta^\sigma P_t^\sigma$

(iii)  $P_t^\sigma P_s^\sigma = P_{t+s}^\sigma$  for any  $t, s \geq 0$ .

Proof: (e)  $\Rightarrow$  (ii), (d)  $\Rightarrow$  (iii), (f) with a modification  $\Rightarrow$  (i).  $\square$

We also collect special basic properties of  $P_t := e^{t\Delta}$ :

Prop II.10.9: (i) All matrix entries of  $P_t$  are ~~real~~ and nonnegative.

(ii)  $P_t c = c$  for any constant function  $c$ .

Proof: Recall all off-diagonal entries of  $\Delta$  are  $\geq 0$ .  
( $-\Delta$  is a generalized Laplacian)

Hence,  $\exists C > 0$  s.t.  $\Delta + C \cdot I$  is entry-wise  $\geq 0$ .

Then  $e^{\Delta + C \cdot I}$  is also entry-wise  $\geq 0$ .

This tells  $P_t = e^{\Delta + C \cdot I} \cdot e^{-C \cdot I}$  is entry-wise  $\geq 0$ .

For a constant function  $c$ , we have  $\Delta c = 0$ .

Hence  $\frac{\partial}{\partial t} P_t c = \Delta c = 0$ . This means  $P_t c = P_0 c = c$ .  $\square$

Rmk: Prop. II.10.9 tells that for any function  $f: V \rightarrow \mathbb{R}$

satisfying  $0 \leq f(x) \leq c$  for all  $x \in V$ , we have  $0 \leq P_t f(x) \leq c, \forall x \in V$ .