

Optimal cuts of random geometric graphs

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Can One Hear the Shape of a Drum? (Kac, 1966)

Let $D \subset \mathbb{R}^d$ be a bounded domain (open, connected) with $d = 2$. The fundamental frequencies of a D -shaped membrane are given by: $\lambda_0 = 0 < \lambda_1 < \lambda_2 \dots$ the eigenvalues of the Laplacian $-\Delta$ on D :

$$\Delta u(x, y) := \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2}, \quad u \in C^2(D), \quad u|_{\partial D} \equiv 0$$

Does $(\lambda_1, \lambda_2, \dots)$ determine the shape of D ?

Many other reasons to be interested in these eigenvalues. For example,

λ_1 is the rate of exponential decay of the survival probability for a Brownian motion in D killed when it hits ∂D .

Cheeger's inequality

Let $d \geq 2$, let $D \subset \mathbb{R}^d$ be a bounded domain. Let λ_1 be the first non-zero eigenvalue of $-\Delta$ on D . Then [Cheeger 1970]

$$\lambda_1 \geq \frac{(\text{CHE}(D))^2}{4}$$

where the *Cheeger constant* of D is given by

$$\text{CHE}(D) := \inf \left\{ \frac{|\partial_D A|}{|A|} : A \subset D, 0 < |A| \leq |D|/2 \right\},$$

$|A|$ denotes the volume of A ,

$|\partial_D A|$ denotes the perimeter of A within D ,

i.e. the surface measure of $\overline{A} \cap \overline{D} \setminus \overline{A}$ (where \overline{A} means closure of A).

Parini 2015: Reverse inequality $\lambda_1 < \frac{\pi^2(\text{CHE}(D))^2}{4}$ for $d = 2$, D convex.

Random (weighted) geometric graphs

Let X_1, X_2, \dots be i.i.d. uniform (D).

For $n \in \mathbb{N}$ and $r > 0$, let $G(n, r)$ be the weighted graph on vertex set $V_n := \{X_1, \dots, X_n\}$ with weights

$$W_{xy} := \phi \left(\frac{|x - y|}{r} \right)$$

where $\phi(t) = \mathbf{1}_{[0,1]}(t)$, $t \geq 0$, and $|\cdot|$ is Euclidean.

i.e., connect any two points of V_n at Euclidean distance at most r_n .

[Could also consider non-uniform densities,
and other weight functions ϕ such as $\phi(t) = \exp(-t^2)$]

Cheeger constant of a graph G ($G = (V, W)$)

Also known as the *conductance* of G :

$$\text{CHE}(G) = \min \left\{ \frac{\partial_G(U)}{\text{vol}_G(U)} : U \subset V, 0 < \text{vol}_G(U) \leq 1/2 \right\}$$

where we set

$$\partial_G(U) := \sum_{v \in U} \sum_{w \in V \setminus U} W_{vw}; \quad \text{vol}_G(U) := \frac{\#(U)}{\#(V)}$$

so $\text{vol}_G(U) \in (0, 1)$. The denominator penalizes unbalanced cuts.

[Alternatively could define $\text{vol}(U)$ by counting edges rather than vertices, and/or change the denominator to $\text{vol}_G(U)\text{vol}_G(V \setminus U)$.]

Uses: provides bounds on mixing times of random walk on graph, bounds on graph laplacian; reasonable criterion for optimal cut.

Machine learning

Aim: learn about D from the sample V_n . In particular:

Can we learn about $\text{CHE}(D)$ from $\text{CHE}(G(n, r_n))$, given $(r_n)_{n \geq 1}$?

$$\begin{aligned} & \text{[Recall } \text{CHE}(D) := \inf \left\{ \frac{|\partial_D A|}{|A|} : A \subset D, 0 < |A| \leq |D|/2 \right\} \\ & \text{CHE}(G) = \min \left\{ \frac{\partial_G(U)}{\text{vol}_G(U)} : U \subset V, 0 < \text{vol}_G(U) \leq 1/2 \right\}] \end{aligned}$$

[Raised by Arias-Castro et al. 2012. Could ask similar in manifolds]

Given $U \subset V_n$, we'll use notation

$$\begin{aligned} \partial_n(U) &:= \partial_{G(n, r_n)}(U), \\ \text{vol}_n(U) &:= \text{vol}_{G(n, r_n)}(U) = \#(U)/n. \end{aligned}$$

Assumptions on D and r_n

$D \subset \mathbb{R}^d$ open and connected.

Also assume $|D| = 1$, and that D has a *Lipschitz boundary* ∂D

[this holds e.g. if ∂D is smooth or D is a cube]

Also, assume that $r_n \ll 1$ and (unless stated otherwise) that

$$nr_n^d \gg \log n,$$

where $a_n \ll b_n$ or $b_n \gg a_n$ means $(a_n/b_n) \rightarrow 0$ as $n \rightarrow \infty$.

Note: $\exists c > 0$: if $nr_n^d \leq c \log n$ then G is not connected so $\text{CHE}(G) = 0$.
Need at least $nr_n^d \geq c \log n$ to have any chance of learning anything from $\text{CHE}(G(n, r_n))$. But want r_n small for computational reasons.

Asymptotic upper bound for $\text{CHE}(G)$ in general D

Choose $A \subset D$ to minimize $|\partial_D A|/|A|$ subject to $0 < |A| \leq \frac{1}{2}$.

Let $U_n = V_n \cap A$. By the SLLN, $\text{vol}_n(U_n) \rightarrow |A|$. Also,

$$\begin{aligned}\mathbb{E}[\partial_n(U_n)] &= n^2 \int_A \int_{D \setminus A} \mathbf{1}_{[0, r_n]}(|y - x|) dy dx \\ &\sim |\partial_D A| \sigma n^2 r_n^{d+1},\end{aligned}$$

with $\sigma := (1/2) \int_{\mathbb{R}^d} x_1 \mathbf{1}_{[0,1]}(|x|) dx$. ['Surface tension' of $\phi = \mathbf{1}_{[0,1]}$], at least if $\partial_D A$ is smooth. So assuming $n^{-2} r_n^{-d-1} \partial_n(U_n)$ is concentrated, and using the Strong Law of Large Numbers for $\text{vol}_n(U_n)$, this gives

$$\begin{aligned}\limsup n^{-2} r_n^{-d-1} \text{CHE}(G(n, r_n)) &\leq \limsup n^{-2} r_n^{-d-1} \left(\frac{\partial_n(U_n)}{\text{vol}_n(U_n)} \right) \\ &= \frac{\sigma |\partial_D A|}{|A|} = \sigma \text{CHE}(D)\end{aligned}$$

THEOREM (García Trillos et al. '16; Müller and P.)

$$\begin{aligned} [\text{Recall } \text{CHE}(D) &:= \inf \left\{ \frac{|\partial_D A|}{|A|} : A \subset D, 0 < |A| \leq |D|/2 \right\} \\ \text{CHE}(G) &= \min \left\{ \frac{\partial_G(U)}{\text{vol}_G(U)} : U \subset V(G), 0 < \text{vol}_G(U) \leq 1/2 \right\}] \end{aligned}$$

Under our conditions ($|D| = 1$, ∂D Lipschitz, $r_n \rightarrow 0$, $nr_n^d \gg \log n$), a.s.:

- $n^{-2}r_n^{-d-1}\text{CHE}(G(n, r_n)) \rightarrow \sigma\text{CHE}(D)$. [already shown \leq]
- If $A \subset D$ is the (essentially) unique Cheeger minimizer, i.e. $|A| < 1/2$ and $\frac{|\partial_D A|}{|A|} < \frac{\partial_D A'}{|A'|}$ for all $A' \subset D$ with $|A' \Delta A| \neq 0$, then

$$n^{-1} \sum_{x \in A_n} \delta_x \rightarrow \text{Leb}_d|_A \quad \text{weakly.}$$

- If A is not unique, we still have convergence on a subsequence.

G. Trillos et al. needed the additional condition $nr_n^2 \gg (\log n)^{3/2}$ if $d = 2$.

Differences between the proofs

- García Trillos *et al.*, in their proof, divide D into n cubes of side $n^{-1/d}$.
- They use minimax grid matching results (Leighton and Shor '89, Shor and Yukich '91) to associate each point of V_n with a nearby cube.
- Hence convert discrete set $U_n \subset V_n$ into a union of cubes.
- Grid matchings need $nr_n^2 \geq C(\log n)^{3/2}$ if $d = 2$.
- In the Müller and P. proof, instead divide D into larger cubes (boxes) of side $\gamma_n r_n$, with $\gamma_n \rightarrow 0$ but $n(\gamma_n r_n)^d \gg \log n$.
- Convert U_n into set of boxes, namely boxes containing 'mostly' points of U_n .

References

- Arias-Castro, E., Pelletier, B. and Pudlo, P. (2012) The normalized graph cut and Cheeger constant: from discrete to continuous. *Adv. Appl. Probab.* **44**, 907-937.
- García Trillos, N. and Slepčev, D. (2016) Continuum limit of total variation on point clouds. *Arch. Ration. Mech. Anal.* **220**, 193-241.
- García Trillos, N., Slepčev, D., von Brecht, J., Laurent, T. and Bresson, X. (2016) Consistency of Cheeger and ratio cuts. *Journal of Machine Learning Research.*
- Müller, T. and Penrose, M.D. (2017+) Optimal Cheeger cuts and bisections of random geometric graphs. In preparation.