

# Notes for the week of 4/8/19 - 4/12/19

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Let  $\{x_1, x_2, \dots\}$  be a sequence of points sampled independently from probability measure  $\mathbb{P}$  with density function  $f$ . For each  $n$ , write  $\mathbf{X}_n = \{x_1, \dots, x_n\} \subseteq \mathbb{R}^d$ . Given some  $\lambda, \sigma > 0$ , let

$$\mathcal{U} = \{x : f(x) \geq \lambda\}, \mathcal{C} = \text{one connected component of } \mathcal{U}, \text{ and } \mathcal{C}_\sigma = \mathcal{C} + B(0, \sigma)$$

Write  $\tilde{\mathbf{X}}_n = \mathcal{C}_\sigma[\mathbf{X}_n]$ ,  $\tilde{E}_n = \{(i, j) : x_i, x_j \in \tilde{\mathbf{X}}_n, \|x_i - x_j\|_2 \leq r\}$  and let  $\tilde{G}_{n,r} = (\tilde{\mathbf{X}}_n, \tilde{E}_n)$ . For a set  $S \subseteq \tilde{\mathbf{X}}_n$ , the normalized cut of  $S$  within  $\tilde{G}_{n,r}$  can be defined as

$$\tilde{\Phi}_{n,r}(S) := \frac{\widetilde{\text{cut}}(S)}{\min\{\widetilde{\text{vol}}(S), \widetilde{\text{vol}}(S^c)\}}, \text{ where } \widetilde{\text{cut}}(S) := \text{cut}(S; \tilde{G}_{n,r}), \widetilde{\text{vol}}(S) := \text{vol}(S; \tilde{G}_{n,r})$$

and in this context  $S^c = \tilde{\mathbf{X}}_n \setminus S$  denotes the complement of  $S$  within  $\tilde{G}_{n,r}$ . Then, the *graph conductance profile* over  $\tilde{G}_{n,r}$  is

$$\tilde{\Phi}_{n,r}(t) := \min_{\substack{S \subseteq \tilde{\mathbf{X}}_n : \\ 0 < \tilde{\pi}_n(S) < t}} \tilde{\Phi}_{n,r}(S)$$

where  $\tilde{\pi}_{n,r}(S) = \frac{\widetilde{\text{vol}}(S)}{\widetilde{\text{vol}}(\tilde{\mathbf{X}}_n)}$ . We will prove a lower bound on the graph conductance profile by a continuous analogue.

## 0.1 Continuous conductance

Let  $\nu$  be the Lebesgue measure over Euclidean space  $\mathbb{R}^d$ , and  $B(x, r)$  be a ball of radius  $r$  centered at  $x$ . For  $S \subset \mathbb{R}^d$  a Borel set,

$$\nu_{\mathbb{P}}(S) := \int_S f(x) dx$$

is the weighted volume.

The  $r$ -ball walk over  $\mathcal{C}_\sigma$  is a Markov chain, with transition probability given by

$$\tilde{P}_{\mathbb{P},r}(x; \mathcal{S}) := \frac{\nu_{\mathbb{P}}(\mathcal{S} \cap B(x, r))}{\nu_{\mathbb{P}}(\mathcal{C}_\sigma \cap B(x, r))} \quad (x \in \mathcal{C}_\sigma, \mathcal{S} \in \mathfrak{B}(\mathcal{C}_\sigma))$$

where  $\mathfrak{B}(\mathcal{C}_\sigma)$  is the Borel  $\sigma$ -algebra of  $\mathcal{C}_\sigma$ .

Denote the stationary distribution for this Markov chain by  $\pi_{\mathbb{P},r}$ , which is defined by the relation

$$\int_{\mathcal{C}_\sigma} \tilde{P}_{\mathbb{P},r}(x; \mathcal{S}) d\pi_{\mathbb{P},r}(x) = \pi_{\mathbb{P},r}(\mathcal{S}). \quad (\mathcal{S} \in \mathfrak{B}(\mathcal{C}_\sigma))$$

Letting the *local conductance* be given by

$$\ell_{\mathbb{P},r}(x) := \nu_{\mathbb{P}}(\mathcal{C}_{\sigma} \cap B(x,r)) \quad (x \in \mathcal{C}_{\sigma})$$

a bit of algebra verifies that

$$\pi_{\mathbb{P},r}(\mathcal{S}) = \frac{\int_{\mathcal{S}} \ell_{\mathbb{P},r}(x) f(x) dx}{\int_{\mathcal{C}_{\sigma}} \ell_{\mathbb{P},r}(x) f(x) dx}. \quad (\mathcal{S} \in \mathfrak{B}(\mathcal{C}_{\sigma}))$$

We next introduce the *ergodic flow*,  $\tilde{Q}_{\mathbb{P},r}$ . Let  $\mathcal{S}_1 \cap \mathcal{S}_2 = \mathcal{C}_{\sigma}$  be a partition of  $\mathcal{C}_{\sigma}$ . Then the ergodic flow between  $\mathcal{S}_1$  and  $\mathcal{S}_2$  is given by

$$\tilde{Q}_{\mathbb{P},r}(\mathcal{S}_1, \mathcal{S}_2) := \int_{\mathcal{S}_1} \tilde{P}_{\mathbb{P},r}(x; \mathcal{S}_2) d\pi_{\mathbb{P},r}(x), \quad (\mathcal{S}_1, \mathcal{S}_2 \in \mathfrak{B}(\mathcal{C}_{\sigma}))$$

the (*continuous*) *normalized cut* by

$$\tilde{\Phi}_{\mathbb{P},r}(\mathcal{S}) := \frac{\tilde{Q}_{\mathbb{P},r}(\mathcal{S}_1, \mathcal{S}_2)}{\min \{\pi_{\mathbb{P},r}(\mathcal{S}), \pi_{\mathbb{P},r}(\mathcal{S}^c)\}}, \quad (\mathcal{S} \in \mathfrak{B}(\mathcal{C}_{\sigma}))$$

and the (*continuous*) *conductance profile* by

$$\tilde{\Phi}_{\nu,r}(t) := \min_{\substack{\mathcal{S} \in \mathfrak{B}(\mathcal{C}_{\sigma}) \\ 0 < \pi_{\mathbb{P},r}(\mathcal{S}) \leq t}} \tilde{\Phi}_{\mathbb{P},r}(\mathcal{S}) \quad (0 < t \leq 1/2)$$

where  $\mathcal{S}^c = \mathcal{C}_{\sigma} \setminus \mathcal{S}$ .

To relate the graph and continuous conductance profiles, we introduce mappings between the data  $\tilde{\mathbf{X}}_n$  and the space  $\mathcal{C}_{\sigma}$ .

## 0.2 Transportation maps and $TL^1$ distance.

Let

$$\tilde{\mathbb{P}}(\mathcal{S}) = \frac{\mathbb{P}(\mathcal{S})}{\mathbb{P}(\mathcal{C}_{\sigma})}, \quad \tilde{\mathbb{P}}_n(\mathcal{S}) := \frac{1}{\tilde{n}} \sum_{x_i \in \tilde{\mathbf{X}}_n} \mathbf{1}(x_i \in \mathcal{S}) \quad (\mathcal{S} \in \mathfrak{B}(\mathcal{C}_{\sigma}))$$

be the (empirical) probability measures, conditional on  $x \sim \mathbb{P}$  lying within  $\mathcal{C}_{\sigma}$  (Here  $\mathfrak{B}(\mathcal{C}_{\sigma})$  is the Borel  $\sigma$ -algebra of  $\mathcal{C}_{\sigma}$ ). A Borel map  $T : \mathcal{C}_{\sigma} \rightarrow \tilde{\mathbf{X}}_n$  is said to be a *transportation map* between  $\tilde{\mathbb{P}}$  and  $\tilde{\mathbb{P}}_n$  if for arbitrary  $\mathcal{S} \in \mathfrak{B}(\mathcal{C}_{\sigma})$ ,

$$\tilde{\mathbb{P}}(\mathcal{S}) = \tilde{\mathbb{P}}_n(T(\mathcal{S})).$$

If a sequence of transportation maps  $\{T_n\}_{n \in \mathbb{N}}$  satisfies  $\|\text{Id} - T_n\|_{L^1(\tilde{\mathbb{P}})} = o_P(1)$ , we refer to it as a sequence of *stagnating transportation maps*. Lemma 1 establishes that with probability one, such a sequence of stagnating transportation maps will exist. In fact, under suitable conditions, the convergence happens at rate  $\left(\frac{\log n}{n}\right)^{1/d}$ .

**Lemma 1** (Adaptation of Proposition 5 of [Garcia Trillos 2016](#)). *With probability one, there exists a sequence of transportation maps  $\{T_n\}_{n \in \mathbb{N}}$ ,  $T_n : \mathcal{C}_{\sigma} \rightarrow \tilde{\mathbf{X}}_n$  such that the following statement holds:*

$$\limsup_{n \rightarrow \infty} \frac{\tilde{n}^{1/d} \|\text{Id} - T_n\|_{L^\infty(\tilde{\mathbb{P}})}}{(\log \tilde{n})^{p_d}} \leq C$$

where  $\text{Id}(x) = x$  is the identity mapping over  $\mathcal{C}_{\sigma}$ ,  $C$  is a universal constant and  $p_d = 3/4$  for  $d = 2$  and  $1/d$  for  $d \geq 3$ .

**Definition 0.1.** For a sequence  $\{u_n\}_{n \in \mathbb{N}} \subseteq L^1(\tilde{\mathbb{P}}_n)$  and  $u \in L^1(\tilde{\mathbb{P}})$ , we say that  $\{u_n\}_{n \in \mathbb{N}}$  converges  $TL^1$  to  $u$  if there exists a sequence of stagnating transportation maps  $\{T_n\}_{n \in \mathbb{N}}$  such that

$$d^{TL^1}(u, u_n) := \int_{\mathcal{C}_\sigma} |u(x) - u_n \circ T_n(x)| d\tilde{\mathbb{P}}(x) \xrightarrow{n} 0 \quad (1)$$

and denote it  $u_n \xrightarrow{TL^1} u$ .

*Remark 1.* Note that as written this is not a metric, as  $u$  and  $u_n$  lie in different spaces. Technically, we can resolve this by writing

$$d^{TL^1}((\tilde{\mathbb{P}}, u), (\tilde{\mathbb{P}}_n, u_n)) = \inf_{\pi \in \Gamma(\tilde{\mathbb{P}}, \tilde{\mathbb{P}}_n)} \iint_{\mathcal{C}_\sigma \times \mathcal{C}_\sigma} |x - y| + |f(x) - g(y)| d\pi(x, y) \quad (2)$$

where  $\gamma$  is the space of couplings over the measures  $\tilde{\mathbb{P}}, \tilde{\mathbb{P}}_n$ . However, it can be shown that (2) converges to zero if and only if (1) is satisfied and

$$\tilde{\mathbb{P}}_n \xrightarrow{w} \tilde{\mathbb{P}}.$$

Since this additional condition will be satisfied with probability one, we simplify things by hereafter referring only to the condition in (1). See [Garcia Trillos 15](#) for more details.

## 1 Lower bound graph conductance profile

In order to lower bound the graph conductance profile  $\tilde{\Phi}_{n,r}(\cdot)$ , we will need to split the analysis into two cases based on the size of  $S \subseteq \tilde{\mathbf{X}}_n$ . Define the *graph local spread* to be

$$s(G) := \frac{9}{10} \min_{u \in V} \{\deg(u; G)\} \cdot \min_{u \in V} \{\pi(u; G)\}. \quad (G = (V, E))$$

Theorem 1 states that for subsets of  $\tilde{\mathbf{X}}_n$  with volume at least  $s(G)$ , the graph normalized cut can be uniformly lower bounded by the continuous normalized cut. Sections 2 and 3 contains the proof of Theorem 1 along with other relevant results.

**Theorem 1.** *Let  $\mathcal{C}_\sigma$  satisfy Assumption 1 with respect to Lipschitz constant  $L$  and convex set  $K$  with diameter  $D_K$ . Then,*

$$\liminf_{n \rightarrow \infty} \left\{ \min_{S \in \mathcal{L}(\tilde{G}_{n,r})} \tilde{\Phi}_{n,r}(S) \right\} \geq \frac{\lambda_\sigma^4 r}{\Lambda_\sigma^4 2^{12} D_K L \sqrt{d}}$$

where  $\mathcal{L}(\tilde{G}_{n,r}) = \left\{ S \subseteq \tilde{\mathbf{X}}_n : \tilde{\pi}_{n,r}(S) \geq s(\tilde{G}_{n,r}) \right\}$

Lemma 2 shows that for the remaining small sets, the graph normalized cut is of constant order.

**Lemma 2.** *Let  $G = (V, E)$  be an undirected graph, and  $\mathcal{L}(G) = \{S \subseteq V : \pi(S; G) \geq s(G)\}$ . Then,*

$$\min_{S \notin \mathcal{L}(G)} \Phi(S; G) \geq \frac{1}{10}.$$

*Proof.* Clearly, for any  $u \in S$

$$\text{cut}(\{u\}, S^c; G) \geq \deg(u; G) - |S| \quad (3)$$

Then, since  $\pi(S; G) \leq s(G)$ ,

$$|S| \leq \pi(S; G) \cdot \frac{\text{vol}(V; G)}{\min_{u \in V} \{\deg(u; G)\}} = \frac{\pi(S; G)}{\min_{u \in V} \{\pi(u; G)\}} \leq \frac{9}{10} \min_{u \in V} \{\deg(u; G)\},$$

and therefore by (3), for any  $u \in S$

$$\text{cut}(\{u\}, S^c; G) \geq \deg(u; G) - \frac{9}{10} \min_{u \in V} \{\deg(u; G)\} \geq \frac{1}{10} \deg(u; G)$$

and the statement follows by summing over all  $u \in S$ .  $\square$

Theorem 1 and Lemma 2 together yield Corollary 1, the main result of these notes.

**Corollary 1** (Lower bound on graph conductance profile). *Let  $\mathcal{C}_\sigma$  satisfy Assumption 1 with respect to Lipschitz constant  $L$  and convex set  $K$  with diameter  $D_K$ . Then, with probability one the following asymptotic lower bound holds on the graph conductance function*

$$\liminf_{n \rightarrow \infty} \tilde{\Phi}_{n,r}(t) \geq \min \left\{ \frac{\lambda_\sigma^4 r}{\Lambda_\sigma^4 2^{12} D_K L \sqrt{d}}, \frac{1}{10} \right\} \quad (0 \leq t \leq \frac{1}{2})$$

## 2 Proofs and Supporting Theory

### 2.1 Proof of Theorem 1.

By Lemma 1, with probability one there exists a sequence of stagnating transportation maps from  $\tilde{\mathbb{P}}$  to  $\tilde{\mathbb{P}}_n$ , which we will denote  $\{T_n\}_{n \in \mathbb{N}}$ .

For  $S \subseteq \tilde{\mathbf{X}}_n$ , let  $T_n^{-1}(S) = \{x \in \mathcal{C}_\sigma : T_n(x) \in S\}$  be the preimage of  $T_n$ , and note that  $T_n^{-1}(S^c) = \mathcal{C}_\sigma \setminus T_n^{-1}(S)$ .

Letting

$$\xi_n := \frac{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-}(x) f(x) dx}{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^+}(x) f(x) dx}, \quad \gamma_n(S) := \frac{\min \left\{ \pi_{\mathbb{P}, r_n^-}(T_n^{-1}(S^c)), \pi_{\mathbb{P}, r_n^-}(T_n^{-1}(S)) \right\}}{\min \left\{ \pi_{\mathbb{P}, r_n^+}(T_n^{-1}(S^c)), \pi_{\mathbb{P}, r_n^+}(T_n^{-1}(S)) \right\}}$$

where  $r_n^\pm := r \pm \|\text{Id} - T_n\|_{L^\infty(\tilde{\mathbb{P}})}$ , by Lemma 5 and Corollary 2 we have that for all  $S \subseteq \tilde{\mathbf{X}}_n$ ,

$$\begin{aligned} \tilde{\Phi}_{n,r}(S) &\geq \xi_n \gamma_n(S) \tilde{\Phi}_{\mathbb{P}, r_n^-}(T_n^{-1}(S)) \\ &\geq \xi_n \gamma_n(S) \frac{\lambda_\sigma^4 r_n^-}{2^{12} \Lambda_\sigma^4 D_K L \sqrt{d}}. \end{aligned} \quad (4)$$

By Lemma 6, with probability one

$$\liminf_{n \rightarrow \infty} \xi_n = 1.$$

By Lemma 8, letting  $c$  be any constant satisfying  $c > \frac{9\lambda_\sigma^4 \nu_d r^d}{50\Lambda_\sigma^2}$ , there exists some  $n \in \mathbb{N}$  such that for all  $S \in \mathcal{L}(\tilde{G}_{n,r})$ ,

$$\pi_{\mathbb{P}, r}(T_n^{-1}(S)) \geq c > 0$$

and therefore by Lemma 7

$$\liminf_{n \rightarrow \infty} \left\{ \inf_{S \in \mathcal{L}(\tilde{G}_{n,r})} \gamma_n(S) \right\} = 1.$$

As Lemma 1 implies  $r_n^- \rightarrow r$  with probability one, an application of Slutsky's Theorem to (4) completes the proof.

## 2.2 Graph functionals to continuous functionals.

Lemmas 3 and 4 provide the necessary bounds for the cut and vol functionals in terms of continuous analogues.

**Lemma 3.** *Let  $S \subseteq \tilde{\mathbf{X}}_n$ , and let  $T_n$  be a transportation map between  $\tilde{\mathbb{P}}$  and  $\tilde{\mathbb{P}}_n$ . Then, letting  $\mathcal{S} = \{x \in \mathcal{C}_\sigma : T_n(x) \in S\}$*

$$\frac{1}{\tilde{n}^2} \widetilde{\text{vol}}(S) \leq \frac{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^+}(x) f(x) dx}{\mathbb{P}(\mathcal{C}_\sigma)^2} \pi_{\mathbb{P}, r_n^+}(\mathcal{S})$$

where  $r_n^+ = r + \|\text{Id} - T_n\|_{L^\infty(\tilde{\mathbb{P}})}$ .

*Proof.* Let  $u : \tilde{\mathbf{X}}_n \rightarrow \{0, 1\}$  be the characteristic function for  $S$ , meaning

$$u(x) = \begin{cases} 1, & x \in S \\ 0, & \text{otherwise} \end{cases}$$

Now, we proceed

$$\begin{aligned} \frac{1}{\tilde{n}^2} \widetilde{\text{vol}}(S_n) &= \frac{1}{\tilde{n}^2} \sum_{x_i, x_j \in \tilde{\mathbf{X}}_n} \mathbf{1}(\|x_i - x_j\| \leq r) |u(x_i)| \\ &= \iint_{\mathcal{C}_\sigma \times \mathcal{C}_\sigma} \mathbf{1}(\|x - y\| \leq r) |u(x)| d\tilde{\mathbb{P}}_n(x) d\tilde{\mathbb{P}}_n(y) \\ &= \iint_{\mathcal{C}_\sigma \times \mathcal{C}_\sigma} \mathbf{1}(\|T_n(x) - T_n(y)\| \leq r) |u \circ T_n(x)| d\tilde{\mathbb{P}}(x) d\tilde{\mathbb{P}}(y) \\ &\leq \iint_{\mathcal{C}_\sigma \times \mathcal{C}_\sigma} \mathbf{1}(\|x - y\| \leq r_n^+) |u \circ T_n(x)| d\tilde{\mathbb{P}}(x) d\tilde{\mathbb{P}}(y) \\ &= \int_S \int_{\mathcal{C}_\sigma \cap B(x, r_n^+)} 1 d\tilde{\mathbb{P}}(y) d\tilde{\mathbb{P}}(x) \end{aligned} \tag{5}$$

By definition we have  $\frac{d\tilde{\mathbb{P}}(x)}{d\tilde{\mathbb{P}}(x)} = \mathbb{P}(\mathcal{C}_\sigma)$ . Therefore,

$$\begin{aligned} \int_S \int_{\mathcal{C}_\sigma \cap B(x, r_n^+)} 1 d\tilde{\mathbb{P}}(y) d\tilde{\mathbb{P}}(x) &= \frac{1}{\mathbb{P}(\mathcal{C}_\sigma)^2} \int_S \int_{\mathcal{C}_\sigma \cap B(x, r_n^+)} 1 d\mathbb{P}(y) d\mathbb{P}(x) \\ &= \frac{1}{\mathbb{P}(\mathcal{C}_\sigma)^2} \int_S \ell_{\mathbb{P}, r_n^+}(x) f(x) dx \\ &= \frac{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r}(x) f(x) dx}{\mathbb{P}(\mathcal{C}_\sigma)^2} \pi_{\mathbb{P}, r_n^+}(\mathcal{S}) \end{aligned}$$

which is the desired upper bound. The lower bound follows a similar proof, with the only change being (5), where  $r_n^+$  is replaced by  $r_n^-$  and the inequality is reversed.  $\square$

**Lemma 4.** *Let  $S \subseteq \tilde{\mathbf{X}}_n$ , and let  $T_n$  be a transportation map between  $\tilde{\mathbb{P}}$  and  $\tilde{\mathbb{P}}_n$ . Then, letting  $\mathcal{S} = \{x \in \mathcal{C}_\sigma : T_n(x) \in S\}$ ,*

$$\frac{1}{\tilde{n}^2} \widetilde{\text{cut}}(S) \geq \frac{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-} f(x) dx}{\mathbb{P}(\mathcal{C}_\sigma)^2} \tilde{Q}_{\mathbb{P}, r_n^-}(\mathcal{S}, \mathcal{S}^c)$$

where  $r_n^- = r - \|\text{Id} - T_n\|_{L^\infty(\tilde{\mathbb{P}})}$

*Proof.* Let  $u : \tilde{\mathbf{X}}_n \rightarrow \{0, 1\}$  be the characteristic function for  $S$ , meaning

$$u(x) = \begin{cases} 1, & x \in S \\ 0, & \text{otherwise} \end{cases}$$

We proceed according to a very similar set of steps as Lemma 3:

$$\begin{aligned} \frac{1}{\tilde{n}^2} \widetilde{\text{cut}}(S) &= \frac{1}{\tilde{n}^2} \sum_{x_i, x_j \in \tilde{\mathbf{X}}_n} \mathbf{1}(\|x_i - x_j\| \leq r) |u(x_i) - u(x_j)| \\ &= \iint_{\mathcal{C}_\sigma \times \mathcal{C}_\sigma} \mathbf{1}(\|x - y\| \leq r) |u(x) - u(y)| d\tilde{\mathbb{P}}_n(x) d\tilde{\mathbb{P}}_n(y) \\ &= \iint_{\mathcal{C}_\sigma \times \mathcal{C}_\sigma} \mathbf{1}(\|T_n(x) - T_n(y)\| \leq r) |u \circ T_n(x) - u \circ T_n(y)| d\tilde{\mathbb{P}}(x) d\tilde{\mathbb{P}}(y) \\ &\geq \iint_{\mathcal{C}_\sigma \times \mathcal{C}_\sigma} \mathbf{1}(\|x - y\| \leq r_n^-) |u \circ T_n(x) - u \circ T_n(y)| d\tilde{\mathbb{P}}(x) d\tilde{\mathbb{P}}(y) \\ &= \int_S \int_{S^c \cap B(x, r_n^-)} d\tilde{\mathbb{P}}(y) d\tilde{\mathbb{P}}(x) \end{aligned}$$

We conclude similarly to the proof of Lemma 3,

$$\begin{aligned} \int_S \int_{S^c \cap B(x, r_n^-)} d\tilde{\mathbb{P}}(y) d\tilde{\mathbb{P}}(x) &= \frac{1}{\mathbb{P}(\mathcal{C}_\sigma)^2} \int_S \int_{S^c \cap B(x, r_n^-)} d\mathbb{P}(y) d\mathbb{P}(x) \\ &= \frac{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-}(x) f(x) dx}{\mathbb{P}(\mathcal{C}_\sigma)^2} \tilde{Q}_{\mathbb{P}, r_n^-}(\mathcal{S}, \mathcal{S}^c). \end{aligned}$$

□

**Lemma 5.** Let  $\{T_n\}_{n \in \mathbb{N}}$  be a sequence of transportation maps from  $\tilde{\mathbb{P}}$  to  $\tilde{\mathbb{P}}_n$ , and let

$$r_n^- = r - \|\text{Id} - T_n\|_{L^\infty(\tilde{\mathbb{P}})}, \quad r_n^+ = r + \|\text{Id} - T_n\|_{L^\infty(\tilde{\mathbb{P}})}.$$

Fix  $S \subseteq \tilde{\mathbf{X}}_n$ . Then, letting  $\mathcal{S} = \{x \in \mathcal{C}_\sigma : T_n(x) \in S\}$ ,

$$\tilde{\Phi}_{n,r}(S) \geq \frac{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-}(x) f(x) dx}{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^+}(x) f(x) dx} \frac{\min \left\{ \pi_{\mathbb{P}, r_n^-}(\mathcal{S}), \pi_{\mathbb{P}, r_n^-}(\mathcal{S}^c) \right\}}{\min \left\{ \pi_{\mathbb{P}, r_n^+}(\mathcal{S}), \pi_{\mathbb{P}, r_n^+}(\mathcal{S}^c) \right\}} \tilde{\Phi}_{\mathbb{P}, r_n^-}(\mathcal{S}) \quad (6)$$

*Proof.* By Lemmas 3 and 4,

$$\frac{\widetilde{\text{cut}}(S)}{\widetilde{\text{vol}}(S)} \geq \frac{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-}(x) f(x) dx}{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^+}(x) f(x) dx} \frac{\tilde{Q}_{\mathbb{P}, r_n^-}(\mathcal{S}, \mathcal{S}^c)}{\pi_{\mathbb{P}, r_n^+}(\mathcal{S})}$$

But, noting that  $\mathcal{S}^c = \{x \in \mathcal{C}_\sigma : T_n(x) \in S^c\}$ , Lemmas 3 and 4 also imply

$$\frac{\widetilde{\text{cut}}(S^c)}{\widetilde{\text{vol}}(S^c)} \geq \frac{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-}(x) f(x) dx}{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^+}(x) f(x) dx} \frac{\tilde{Q}_{\mathbb{P}, r_n^-}(\mathcal{S}^c, \mathcal{S})}{\pi_{\mathbb{P}, r_n^+}(\mathcal{S}^c)}$$

and as  $\tilde{Q}_{\mathbb{P}, r_n^-}(\cdot, \cdot)$  is symmetric in its arguments we obtain

$$\frac{\widetilde{\text{cut}}(S)}{\min \left\{ \widetilde{\text{vol}}(S), \widetilde{\text{vol}}(S^c) \right\}} \geq \frac{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-}(x) f(x) dx}{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^+}(x) f(x) dx} \frac{\tilde{Q}_{\mathbb{P}, r_n^-}(\mathcal{S}, \mathcal{S}^c)}{\min \left\{ \pi_{\mathbb{P}, r_n^+}(\mathcal{S}), \pi_{\mathbb{P}, r_n^+}(\mathcal{S}^c) \right\}},$$

and the proof is complete. □

### 2.3 Perturbation asymptotics.

In light of Lemma 1, the error incurred in (6) by the use of  $r_n^+$  and  $r_n^-$  as opposed to  $r$  is asymptotically negligible.

**Lemma 6** (Continuity of local conductance). *Letting  $\{T_n\}_{n \in \mathbb{N}}$  be a sequence of stagnating transportation maps, and  $r_n^\pm = r \pm \|\text{Id} - T_n\|_{L^\infty(\tilde{\mathbb{P}})}$ , with probability one the following holds:*

$$\limsup_{n \rightarrow \infty} \frac{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^+}(x) f(x) dx}{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-}(x) f(x) dx} = 1$$

*Proof.* Let  $\mathcal{R}_n(x) := \{x' \in \mathcal{C}_\sigma : x' \in B(x, r_n^+), x' \notin B(x, r_n^-)\}$ , we have

$$\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^+}(x) f(x) dx = \int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-}(x) f(x) dx + \int_{\mathcal{C}_\sigma} \int_{\mathcal{R}_n} f(y) f(x) dy dx.$$

and therefore

$$\frac{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^+}(x) f(x) dx}{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-}(x) f(x) dx} = 1 + \frac{\int_{\mathcal{C}_\sigma} \int_{\mathcal{R}_n} f(y) f(x) dy dx}{\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-}(x) f(x) dx}.$$

We upper bound the remainder term

$$\int_{\mathcal{C}_\sigma} \int_{\mathcal{R}_n} f(y) f(x) dy dx \leq P(\mathcal{C}_\sigma) \Lambda_\sigma((r_n^+)^d - r^d) \nu^d$$

and taking limits as  $n \rightarrow \infty$  we obtain with probability one

$$\limsup_{n \rightarrow \infty} \int_{\mathcal{C}_\sigma} \int_{\mathcal{R}_n} f(y) f(x) dy dx \leq \limsup_{n \rightarrow \infty} P(\mathcal{C}_\sigma) \Lambda_\sigma((r_n^+)^d - r^d) \nu^d = 0$$

by the stagnating property of  $\{T_n\}_{n \in \mathbb{N}}$ .

We apply a similar analysis to the denominator.

$$\int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-}(x) f(x) dx = \int_{\mathcal{C}_\sigma} \int_{B(x, r_n^-)} f(y) f(x) dy dx \geq \frac{6}{25} \mathbb{P}(\mathcal{C}_\sigma) \lambda_\sigma(r_n^-)^d \nu_d$$

and therefore by the stagnating property of  $\{T_n\}_{n \in \mathbb{N}}$  and Lemma 9,

$$\liminf_{n \rightarrow \infty} \int_{\mathcal{C}_\sigma} \ell_{\mathbb{P}, r_n^-}(x) f(x) dx = \frac{6}{25} \mathbb{P}(\mathcal{C}_\sigma) \lambda_\sigma^2 r^d \nu_d > 0$$

again with probability one.

The desired result then follows from an application of Slutsky's Theorem.  $\square$

**Lemma 7** (Continuity of stationary distribution). *Let  $c > 0$  be a fixed constant,  $\{T_n\}_{n \in \mathbb{N}}$  be a sequence of stagnating transportation maps, and  $r_n^\pm = r \pm \|\text{Id} - T_n\|_{L^\infty(\tilde{\mathbb{P}})}$ . With probability one the following statement holds:*

$$\liminf_{n \rightarrow \infty} \frac{\min \left\{ \pi_{\mathbb{P}, r_n^-}(\mathcal{S}), \pi_{\mathbb{P}, r_n^-}(\mathcal{S}^c) \right\}}{\min \left\{ \pi_{\mathbb{P}, r_n^+}(\mathcal{S}), \pi_{\mathbb{P}, r_n^+}(\mathcal{S}^c) \right\}} = 1$$

*uniformly over all sets  $\mathcal{S} \subseteq \mathfrak{B}(\mathcal{C}_\sigma)$  satisfying  $\min \{ \pi_{\mathbb{P}, r}(\mathcal{S}), \pi_{\mathbb{P}, r}(\mathcal{S}^c) \} > c$ .*

*Proof.* It will be sufficient to show that

$$\liminf_{n \rightarrow \infty} \frac{\pi_{\mathbb{P}, r_n^-}(\mathcal{S})}{\pi_{\mathbb{P}, r_n^+}(\mathcal{S})} \text{ and } \liminf_{n \rightarrow \infty} \frac{\pi_{\mathbb{P}, r_n^-}(\mathcal{S}^c)}{\pi_{\mathbb{P}, r_n^+}(\mathcal{S}^c)} = 1.$$

and we will show only that  $\liminf_{n \rightarrow \infty} \frac{\pi_{\mathbb{P}, r_n^-}(\mathcal{S})}{\pi_{\mathbb{P}, r_n^+}(\mathcal{S})} = 1$ . The result for  $\mathcal{S}^c$  is identical.

The proof proceeds similarly to Lemma 6. Letting

$$\mathcal{R}_n(x) := \{x' \in \mathcal{S} : x' \in B(x, r_n^+), x' \notin B(x, r_n^-)\}$$

Rewriting

$$\frac{\pi_{\mathbb{P}, r_n^-}(\mathcal{S})}{\pi_{\mathbb{P}, r_n^+}(\mathcal{S})} = 1 - \frac{\int_{\mathcal{S}} \int_{\mathcal{R}_n} f(y) f(x) dy dx}{\pi_{\mathbb{P}, r_n^+}(\mathcal{S})}$$

we have that

$$\liminf_{n \rightarrow \infty} \int_{\mathcal{S}} \int_{\mathcal{R}_n} f(y) f(x) dy dx \leq \mathbb{P}(\mathcal{S}) \Lambda_{\sigma} \liminf_{n \rightarrow \infty} ((r_n^+)^d - (r_n^-)^d) \nu^d = 0$$

where the equality occurs with probability one. On the other hand by hypothesis

$$\limsup_{n \rightarrow \infty} \pi_{\mathbb{P}, r_n^+}(\mathcal{S}) \geq c > 0.$$

and the result follows by Slutsky's Theorem.  $\square$

**Lemma 8** (Stationary distribution lower bound). *With probability one, the following statement holds: let  $\{T_n\}_{n \in \mathbb{N}}$  be a sequence of stagnating transportation maps from  $\mathbb{P}$  to  $\mathbb{P}_n$ . Then, for any  $\epsilon > 0$ , there exists some  $m \in \mathbb{N}$  such that for all  $n \geq m$ ,*

$$\min_{S \in \mathcal{L}(\tilde{G}_{n,r})} \pi_{\mathbb{P}, r}(T_n^{-1}(S)) \geq \frac{9\lambda_{\sigma}^4 \nu_d r^d}{50\Lambda_{\sigma}^2} - \epsilon$$

*Proof.* Fix  $\epsilon > 0$ , and let  $S \in \mathcal{L}(\tilde{G}_{n,r})$  be arbitrary. Write  $\mathcal{S} := T_n^{-1}(S)$ .

We can upper bound  $\tilde{\pi}_{n,r}(S)$  by  $\pi_{\mathbb{P}, r}(\mathcal{S})$  plus a remainder term.

$$\begin{aligned} \tilde{\pi}_{n,r}(S) &\leq \frac{\int_{\mathcal{S}} \int_{\mathcal{C}_{\sigma}} \mathbf{1}(\|x - x'\| \leq r_n^+) f(x') f(x) dx' dx}{\int_{\mathcal{C}_{\sigma}} \int_{\mathcal{C}_{\sigma}} \mathbf{1}(\|x - x'\| \leq r_n^+) f(x') f(x) dx' dx} \\ &= \pi_{\mathbb{P}, r_n^-}(\mathcal{S}) + \frac{\int_{\mathcal{S}} \int_{\mathcal{C}_{\sigma}} \mathbf{1}(r_n^- \leq \|x - x'\| \leq r_n^+) f(x') f(x) dx' dx}{\int_{\mathcal{C}_{\sigma}} \int_{\mathcal{C}_{\sigma}} \mathbf{1}(\|x - x'\| \leq r_n^+) f(x') f(x) dx' dx} \end{aligned} \quad (7)$$

Clearly  $\pi_{\mathbb{P}, r_n^-}(\mathcal{S}) \leq \pi_{\mathbb{P}, r}(\mathcal{S})$ . Moreover

$$\frac{\int_{\mathcal{S}} \int_{\mathcal{C}_{\sigma}} \mathbf{1}(r_n^- \leq \|x - x'\| \leq r_n^+) f(x') f(x) dx' dx}{\int_{\mathcal{C}_{\sigma}} \int_{\mathcal{C}_{\sigma}} \mathbf{1}(\|x - x'\| \leq r_n^+) f(x') f(x) dx' dx} \leq \frac{\Lambda_{\sigma}}{\lambda_{\sigma}} \left( \frac{r_n^+ - r_n^-}{r_n^-} \right)^d$$

and by (7), we have

$$s(\tilde{G}_{n,r}) \leq \tilde{\pi}_{n,r}(S) \leq \pi_{\mathbb{P}, r}(\mathcal{S}) + \frac{\Lambda_{\sigma}}{\lambda_{\sigma}} \left( \frac{r_n^+ - r_n^-}{r_n^-} \right)^d$$

The remainder term  $\frac{\Lambda_{\sigma}}{\lambda_{\sigma}} \left( \frac{r_n^+ - r_n^-}{r_n^-} \right)^d$  is independent of  $S$  and asymptotically  $o_p(1)$  by Lemma 1. An application of Lemma 11 completes the proof.  $\square$



## 2.4 Continuous conductance function.

In Theorem 2, we restate a necessary result from the 3/20 weekly notes.

**Assumption 1** (Embedding). *Assume there exists  $K \subset \mathbb{R}^d$  convex space, mapping  $g : K \rightarrow \mathcal{C}_\sigma$ , and constant  $L < \infty$  such that*

$$\forall x, y \in K, |g(x) - g(y)| \leq L |x - y|, \text{ and } \det(D_x g) = 1.$$

*In other words,  $g$  is measure-preserving and  $L$ -Lipschitz.*

**Theorem 2.** *Assume  $\mathcal{C}_\sigma \subset \mathbb{R}^d$  satisfies Assumption 1 with respect to some convex set  $K \subset \mathbb{R}^d$  and Lipschitz function  $g$  with Lipschitz constant  $L < \infty$ . Then, for any  $0 < r < \sigma/2\sqrt{d}$ , the continuous conductance function of the speedy  $r$ -ball walk satisfies*

$$\tilde{\Phi}_{\nu,r}(t) \geq \frac{r}{2^{12} D_K L \sqrt{d}}.$$

Corollary 2 follows almost immediately for Theorem 2.

**Corollary 2.** *Assume  $\mathcal{C}_\sigma \subset \mathbb{R}^d$  satisfies Assumption 1 with respect to some convex set  $K \subset \mathbb{R}^d$  and Lipschitz function  $g$  with Lipschitz constant  $L < \infty$ . Then, for any  $0 < r < \sigma/2\sqrt{d}$ , the continuous conductance function of the speedy  $r$ -ball walk satisfies*

$$\tilde{\Phi}_{\mathbb{P},r}(t) \geq \frac{\lambda_\sigma^4 r}{2^{12} \Lambda_\sigma^4 D_K L \sqrt{d}}.$$

where we recall  $\lambda_\sigma = \inf_{x \in \mathcal{C}_\sigma} f(x)$  and  $\Lambda_\sigma = \sup_{x \in \mathcal{C}_\sigma} f(x)$ .

## 3 Other results

We state Lemma 9 without proof. The proof in the uniform case can be found in the 3/20 notes.

**Lemma 9.** *Let  $x \in \mathcal{C}_\sigma$ . Then, for any  $r < \frac{\sigma}{2\sqrt{d}}$ ,*

$$\ell_{\mathbb{P},r}(x) \geq \frac{6\lambda_\sigma^2}{25} r^d \nu_d$$

and for any  $r > 0$ ,

$$\ell_{\mathbb{P},r}(x) \leq \Lambda_\sigma^2 r^d \nu_d$$

**Lemma 10.** *Let*

$$\mu' = \frac{6(n-1)\lambda_\sigma^2}{25} r^d \nu_d, \quad \mu = (n-1)\Lambda_\sigma^2 r^d \nu_d.$$

*Then for any  $\delta \in [0, 1]$ ,*

$$\begin{aligned} \Pr\left(\min_{x_i \in \tilde{\mathbf{X}}_n} \deg(x_i; \tilde{G}_{n,r}) \leq (1-\delta)\mu'\right) &\leq (n-1) \exp\{-\delta^2 \mu'/2\} \\ \Pr\left(\max_{x_i \in \tilde{\mathbf{X}}_n} \deg(x_i; \tilde{G}_{n,r}) \geq (1+\delta)\mu\right) &\leq (n-1) \exp\{-\delta^2 \mu/3\} \end{aligned}$$

*Proof.* For each  $x_i \in \tilde{\mathbf{X}}_n$ , letting  $Y_{ij} = \mathbf{1}(x_j \in \mathcal{C}_\sigma, \|x_i - x_j\| \leq r)$  we can write  $\deg(x_i; \tilde{G}_{n,r}) = \sum_{j \neq i} Y_{ij}$ . Note that  $\mathbb{E}(Y_{ij}|x_i) = \ell_{\mathbb{P},r}(x_i)$ , and  $Y_{ij}$  and  $Y_{ij'}$  are independent for all  $j \neq j'$ . Therefore the desired result follows from Lemmas 9 and 12 along with a union bound.  $\square$

**Lemma 11.** *With probability one, the following statement holds:*

$$\liminf_{n \rightarrow \infty} s(\tilde{G}_{n,r}) \geq \frac{9\lambda_\sigma^4 \nu_d r^d}{50\Lambda_\sigma^2}$$

*Proof.* We rewrite

$$\begin{aligned} s(\tilde{G}_{n,r}) &= \frac{9 \left[ \min_{x \in \tilde{\mathbf{X}}_n} \left\{ \widetilde{\deg_{n,r}}(x) \right\} \right]^2}{\widetilde{\text{vol}}_{n,r}(\tilde{\mathbf{X}}_n)} \\ &\geq \frac{9 \left[ \min_{x \in \tilde{\mathbf{X}}_n} \left\{ \widetilde{\deg_{n,r}}(x) \right\} \right]^2}{n \max_{x \in \tilde{\mathbf{X}}_n} \left\{ \widetilde{\deg_{n,r}}(x) \right\}} \end{aligned}$$

The statement follows by Lemma 10 and the Borel-Cantelli Lemma.  $\square$

**Lemma 12** (Multiplicative Chernoff Bound). *Let  $p', p \in [0, 1]$ , and let  $Y_1, \dots, Y_n$  be independent  $\{0, 1\}$ -valued random variables with  $p' \leq E(Y_i) \leq p$  for all  $i = 1, \dots, n$ . Then, letting  $\mu = pn$  and  $\mu' = p'n$ ,*

$$\begin{aligned} \Pr\left(\sum_{i=1}^n Y_i \leq (1 - \delta)\mu'\right) &\leq \exp\{-\delta^2 \mu' / 2\} \\ \Pr\left(\sum_{i=1}^n Y_i \geq (1 + \delta)\mu\right) &\leq \exp\{-\delta^2 \mu / 3\} \end{aligned}$$

for any  $\delta \in [0, 1]$ .