

Lipschitz and outer bi-Lipschitz extendability

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Plan

- The Lipschitz extendability problem
- Known results and open problems
- Vertex sparsifiers
- Connection between vertex sparsifiers and the Lipschitz extendability problem
- Outer bi-Lipschitz extendability: definition, results, and open problems

Hahn–Banach Theorem



H. Hahn



S. Banach

Let V be a normed space and $L \subset V$ be its linear subspace. Every bounded linear map

$$f: L \rightarrow \mathbb{R}$$

can be extended to $\tilde{f}: V \rightarrow \mathbb{R}$ so that

$$\|\tilde{f}\| = \|f\|$$

Is there an analogue for

- Lipschitz maps?
- maps into \mathbb{R}^d or other normed spaces?

Preliminaries

$f: X \rightarrow Y$ is Lipschitz if for every $u, v \in X$

$$d_Y(f(u), f(v)) \leq C d_X(u, v)$$

The Lipschitz constant $\|f\|_{Lip}$ of f is the minimum C s.t. that the inequality holds.

f is bi-Lipschitz if for some $C_1, C_2 > 0$, every $u, v \in X$

$$C_1 d_X(u, v) \leq d_Y(f(u), f(v)) \leq C_2 d_X(u, v)$$

The bi-Lipschitz constant or distortion $D(f)$ of f is the minimum of C_2/C_1 s.t. that the inequality holds.

Preliminaries

ℓ_p^d is \mathbb{R}^d equipped with the $\|\cdot\|_p$ norm:

$$\|x\|_p = \left(\sum |x_i|^p \right)^{1/p}$$

$$\|x\|_\infty = \max |x_i|$$

McShane–Whitney Theorem

“non-linear Hahn–Banach”



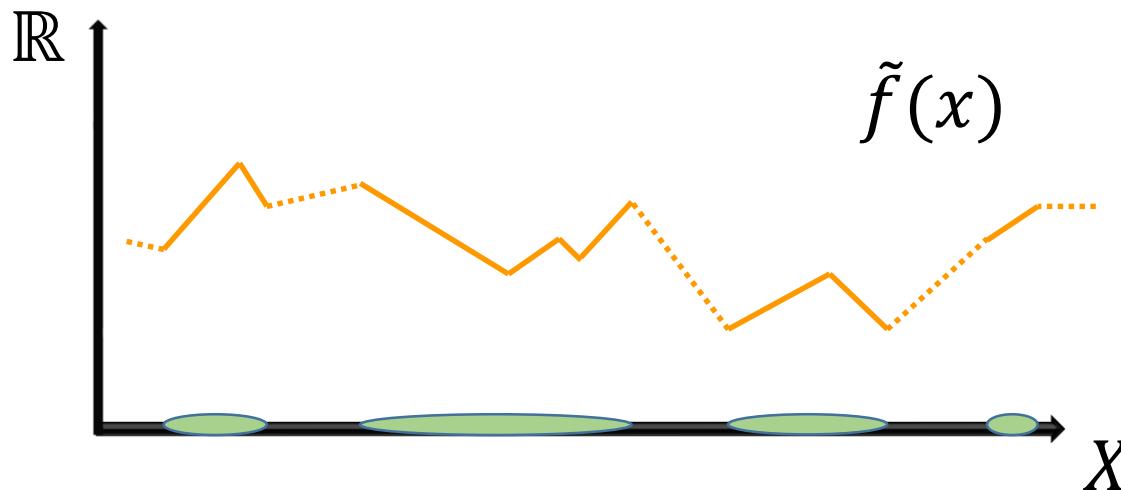
E. McShane



H. Whitney

Let (X, d) be a metric space and $A \subset X$. Every Lipschitz map $f: A \rightarrow \mathbb{R}$ can be extended to $\tilde{f}: X \rightarrow \mathbb{R}$ so that

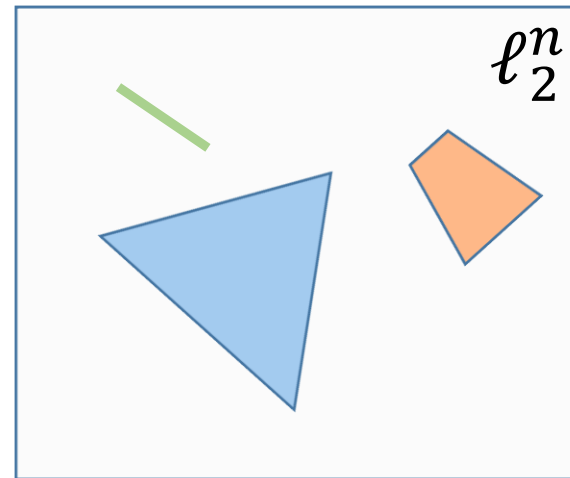
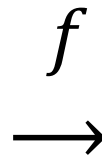
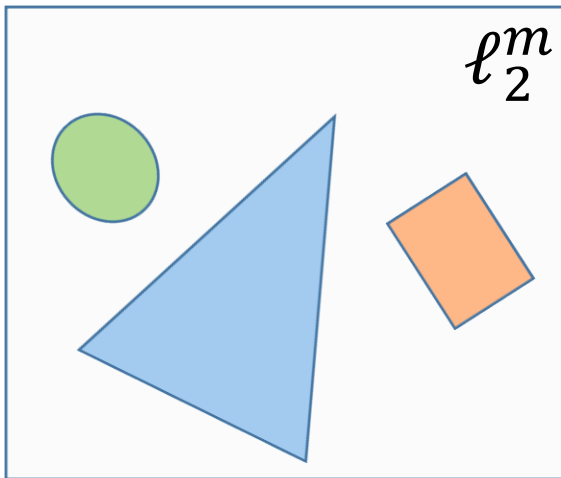
$$\|\tilde{f}\|_{Lip} = \|f\|_{Lip}$$



Kirszbraun Theorem

Let $A \subset \ell_2^m$. Every Lipschitz map $f: A \rightarrow \ell_2^n$ can be extended to $\tilde{f}: \ell_2^m \rightarrow \ell_2^n$ so that

$$\|\tilde{f}\|_{Lip} = \|f\|_{Lip}$$



Lipschitz Extension Constant

Let X be a metric space and Y be a normed space.

$e_k(X, Y)$ is the min C s.t. for every $A \subset X$ of size $\leq k$ and

$$f: A \rightarrow Y$$

there exists an extension $\tilde{f}: X \rightarrow Y$ with

$$\|\tilde{f}\|_{Lip} \leq C \|f\|_{Lip}$$

McShane–Whitney: $e_k(X, \mathbb{R}) = 1$

Kirszbraun: $e_k(\ell_2, \ell_2) = 1$

Lipschitz Extension Constant

In general, $e_k(X, Y) > 1$

E.g., $e_3(\ell_1^3, \ell_2^2) \geq \sqrt{3}$

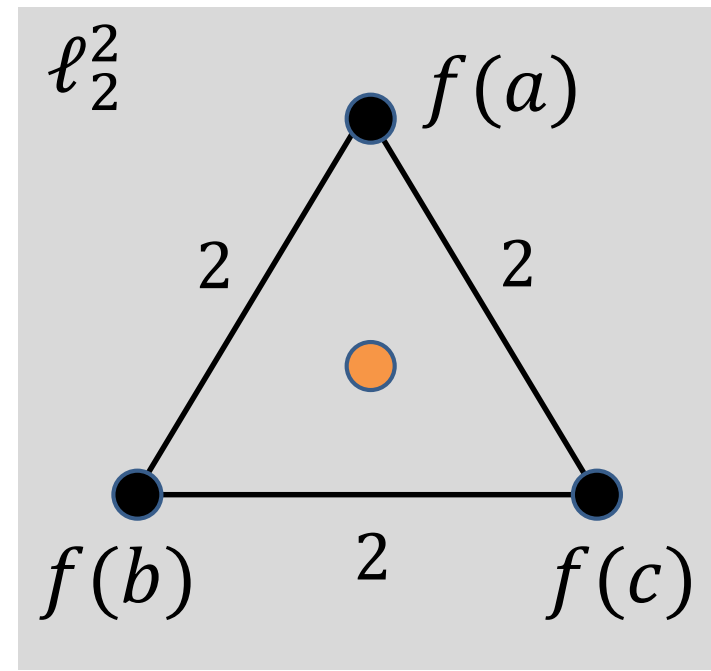
$$A = \{a, b, c\}$$

$$a = (1, 0, 0)$$

$$b = (0, 1, 0)$$

$$c = (0, 0, 1)$$

$$d = (0, 0, 0)$$



| $X \rightarrow Y$ | $e_k(X, Y)$ | |
|--|--|-------------------------------------|
| any $\rightarrow \mathbb{R}$ or ℓ_∞ | 1 | McShane, Whitney '34 |
| $\ell_2 \rightarrow \ell_2$ | 1 | Kirszbraun '34 |
| $\ell_p \rightarrow \ell_2$ $p \leq 2$ | $\leq C_p (\log k)^{\frac{1}{p} - \frac{1}{2}}$ | Marcus, Pisier '84 |
| $1 < p < 2$ | $\geq c_p \left(\frac{\log k}{\log \log k} \right)^{\frac{1}{p} - \frac{1}{2}}$ | Johnson, Lindenstrauss '84 |
| any $\rightarrow \ell_2$ | $\leq C \sqrt{\log k}$ | JL '84 |
| $\ell_p \rightarrow \ell_q$ $1 < q \leq 2 \leq p$ | $\leq 24 \sqrt{\frac{p-1}{q-1}}$ | Naor, Peres, Schramm, Sheffield '04 |
| other $p, q \in (1, \infty)$ | $e_k \rightarrow \infty$ as $k \rightarrow \infty$ | |

Extension Results

Johnson–Lindenstruass–Schechtman '86

$$e_k(X, Y) \leq C \log k$$

Lee–Naor '03

$$e_k(X, Y) \leq \frac{C \log k}{\log \log k}$$

Best lower bounds are:

$$e_k \gtrsim c \sqrt{\log k}$$

Open Problem: what is the dependence of e_k on k ?

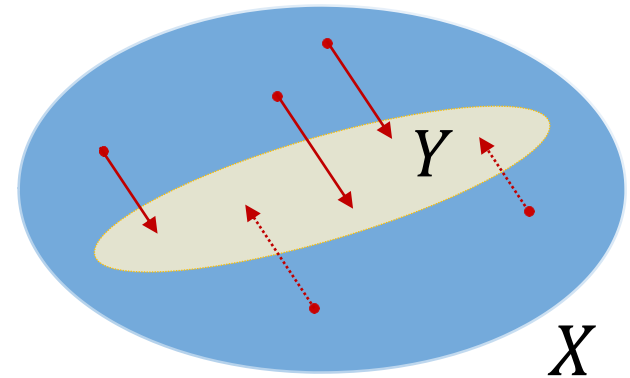
[JL '84] Technique for proving lower bounds on $e_k(X, Y)$

Prove a lower bound for linear extensions

Reinterpret it as a lower bound for Lipschitz extensions

- Linear extension (“projection”) constant is up to $\sqrt{\dim Y}$

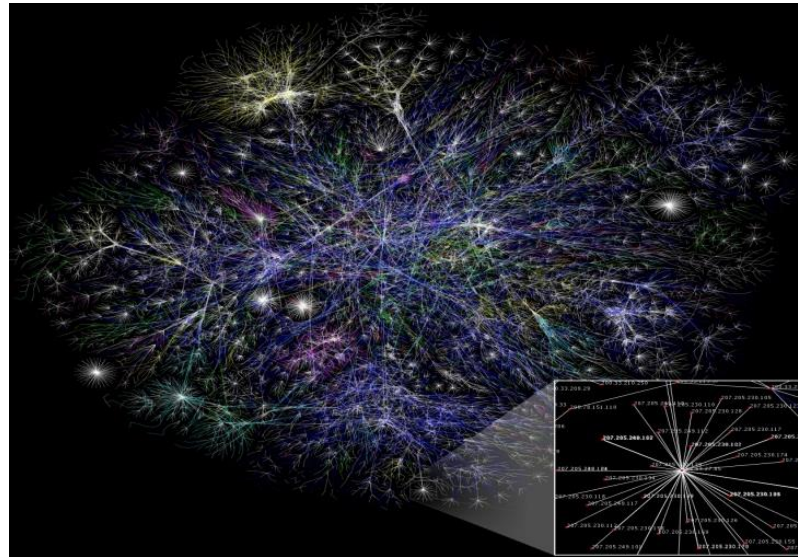
- [JL '84] $e_k(X, Y) \geq c \sqrt{\frac{\log k}{\log \log k}}$.



Open Problems

- Can the upper bound of $\sim \log k / \log \log k$ be improved?
- Are there any X and Y with $e_k(X, Y) \gg \sqrt{\log k}$?
- Ball '92: Is it true that $e_k(\ell_2, \ell_1) \leq C < \infty$?

Graph Sparsification

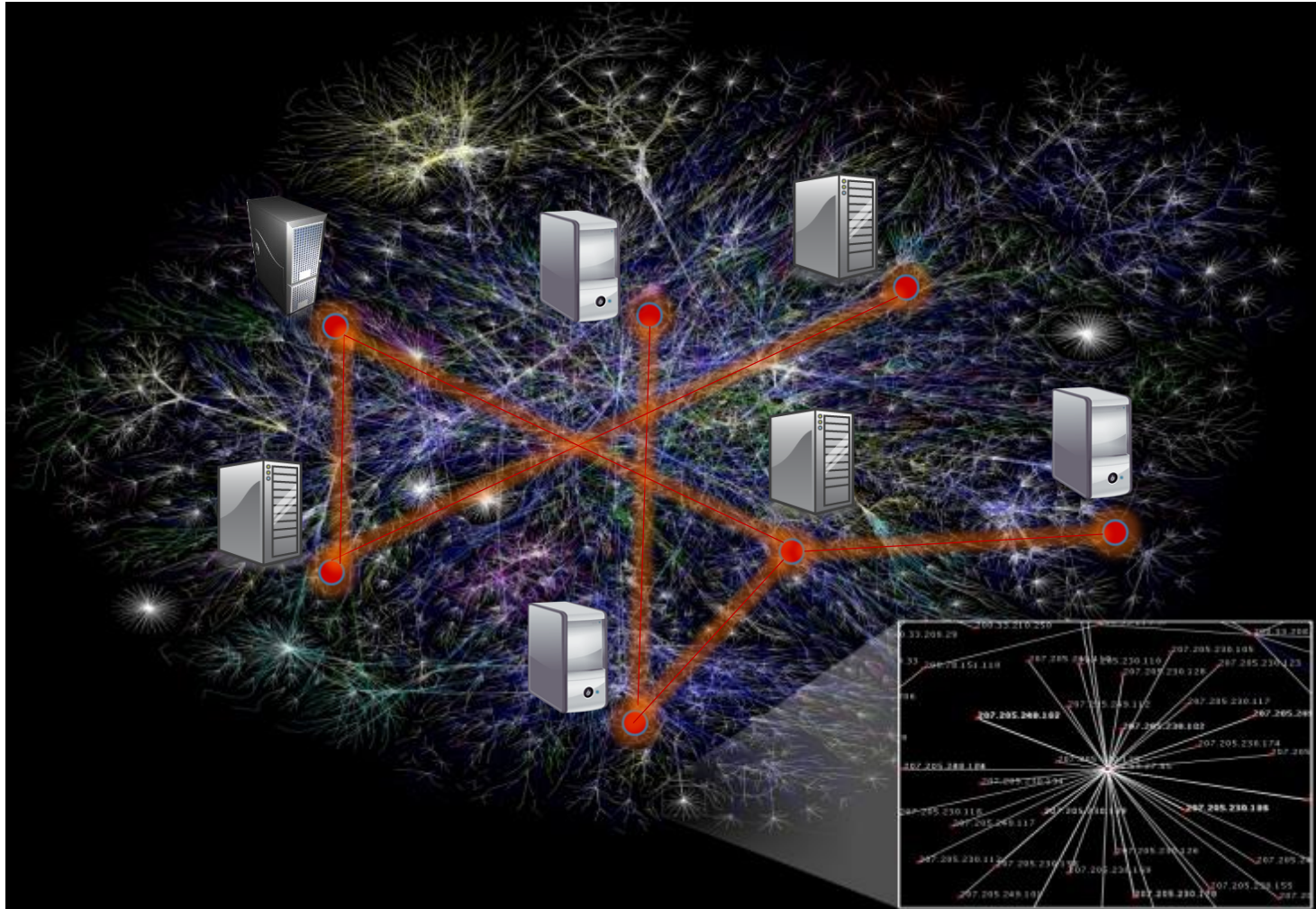


Given: a huge graph G

Goal: find a “simpler” graph H “similar” to G

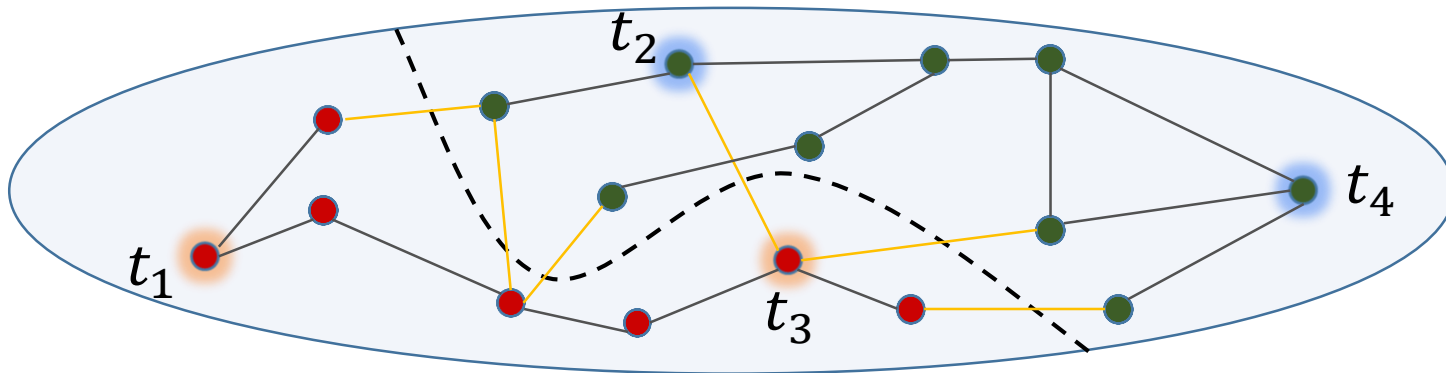
- compact representation
- algorithms work faster on the new graph
- can obtain better approximation results

Bottleneck & Routing Problems



Bottleneck Problem

- graph $G = (V, E)$ with edge capacities c_e
- set of terminals $T \subset V$

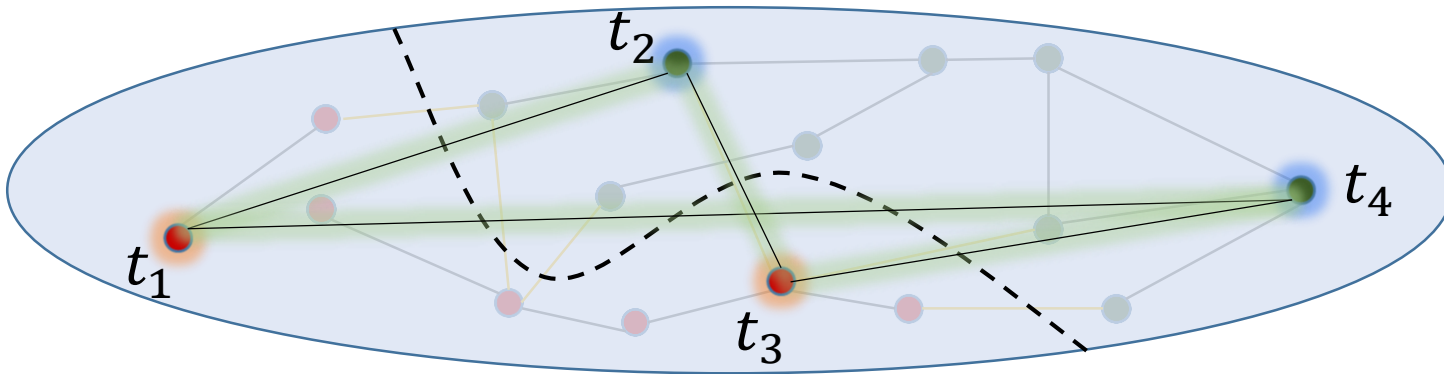


For $S \subset T$, $\text{bk}_G(S)$ is the capacity of the minimum capacity cut in G that separates S and $T \setminus S$ in G .

Bottleneck Problem

[Moitra '09] Graph $H = (T, E')$ with capacities c'_e is a **vertex cut sparsifier** for G with distortion $D \geq 1$ if

$$\text{bk}_G(S) \leq \text{bk}_H(S) \leq D \cdot \text{bk}_G(S) \quad \forall S \subset T$$



Given H , can easily compute bottlenecks between terminals in the network!

Network Routing Problem

Routing problem: send a certain amount of data d_{ij} from each terminal t_i to t_j so that the total amount sent over each edge e is at most its capacity c_e .

[LM '10] A **vertex flow sparsifier** is an analogue of a vertex cut sparsifier for the network routing problem.

Known Results

Moitra '09 and Leighton and Moitra '10

$$k = |T|$$

- $C \log k / \log \log k$ existential upper bound
- $C \log^2 k / \log \log k$ algorithmic upper bound
- $C > 1$ lower bound for cut sparsifiers
- $\Omega(\log \log k)$ lower bound for flow sparsifiers

Open Questions:

- $C \log k / \log \log k$ algorithmic upper bound?
- Better lower bounds?
- Better upper bounds?

Papers on Vertex Sparsification

Charikar, Leighton, Li and Moitra '10

Englert, Gupta, Krauthgamer, Räcke, Talgam
and Talwar '10

Makarychev and Makarychev '10

Main Results

- Define “Metric Sparsifiers”
- Give $C \log k / \log \log k$ algorithmic upper bound [independently, CLLM '10, EGKRTT '10]
- Establish a direct connection between Vertex Sparsifiers and Lipschitz Extendability

$$Q_k^{cut} = e_k(\ell_1, \ell_1)$$
$$Q_k^{flow} = e_k(\ell_\infty, \ell_\infty \oplus_1 \dots \oplus_1 \ell_\infty)$$

Lower bounds via Lipschitz Extendability

Using lower bounds for “projection constants”
[Grünbaum '60], we get

$$Q_k^{flow} \geq e_k(\ell_\infty, \ell_1) \geq C \sqrt{\log k / \log \log k}$$

Figiel, Johnson, and Schechtman '88 implies

$$Q_k^{cut} = e_k(\ell_1, \ell_1) \geq \frac{C \sqrt{\log k}}{\log \log k}$$

Proof Idea: $Q_k^{cut} \leq Q \equiv e_k(\ell_1, \ell_1)$

Consider a game: G and $\{c_e\}$ are fixed

Alice: defines H by providing c'_e

Bob: presents $(S_1, T \setminus S_1)$ and $(S_2, T \setminus S_2)$

$$\text{bk}_G(S_1) \leq \text{bk}_H(S_1) \quad \text{and} \quad \text{bk}_H(S_2) \leq Q \text{bk}_G(S_2)$$

yes

Alice wins

no

Bob wins

Proof Idea: $Q_k^{cut} \leq Q \equiv e_k(\ell_1, \ell_1)$

Consider a game: G and $\{c_e\}$ are fixed

Bob: **distribution** of $(S_1, T \setminus S_1)$ and $(S_2, T \setminus S_2)$
Alice: defines H by providing c'_e

$$\mathbb{E}k_G(S_1) \leq \mathbb{E}k_H(S_1) \quad \mathbb{E}k_H(S_2) \leq Q \quad \mathbb{E}k_G(S_2)$$

yes

Alice wins

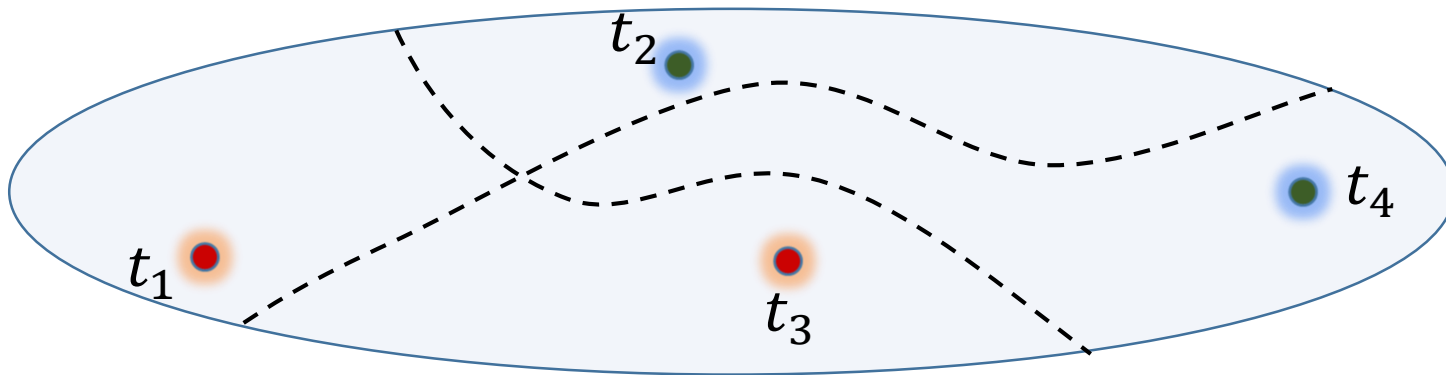
no

Bob wins

Distribution of cuts

Distribution \mathcal{D} of cuts $(S, T \setminus S)$ on T defines a map $f: T \rightarrow L_1(\Omega, \mu)$:

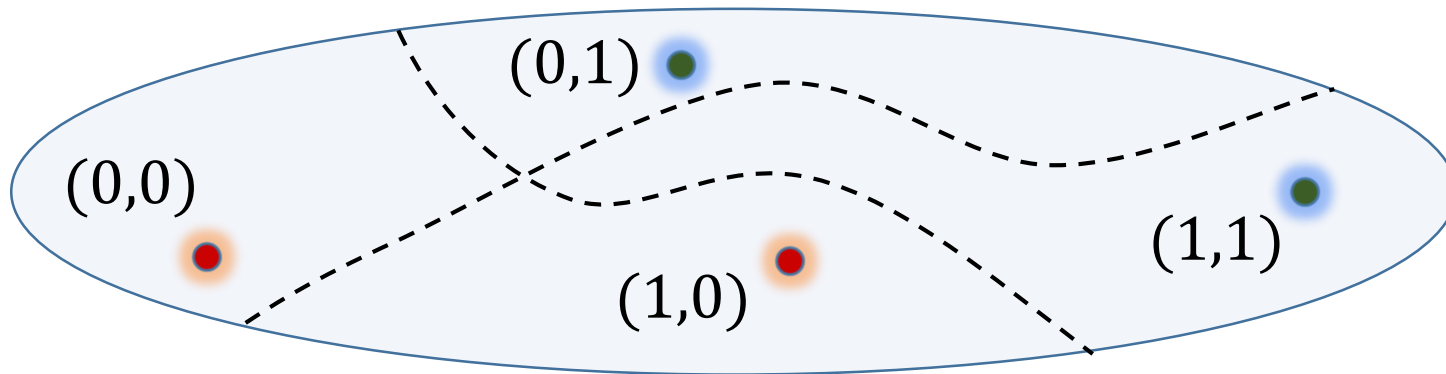
$$f(u) = \begin{cases} 0, & \text{if } x \in S \\ 1, & \text{if } x \notin S \end{cases}$$



Distribution of cuts

Distribution \mathcal{D} of cuts $(S, T \setminus S)$ on T defines a map $f: T \rightarrow L_1(\Omega, \mu)$:

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Distribution of cuts

$$f(x) = \begin{cases} 0, & \text{if } x \in S \\ 1, & \text{if } x \notin S \end{cases}$$

$$\Pr[u, v \text{ are separated by } S] = \|f(u) - f(v)\|_1$$

$$\mathbb{E}[\text{bk}_H(S)] = \sum_{u,v \in T} c'(u, v) \cdot \|f(u) - f(v)\|_1$$

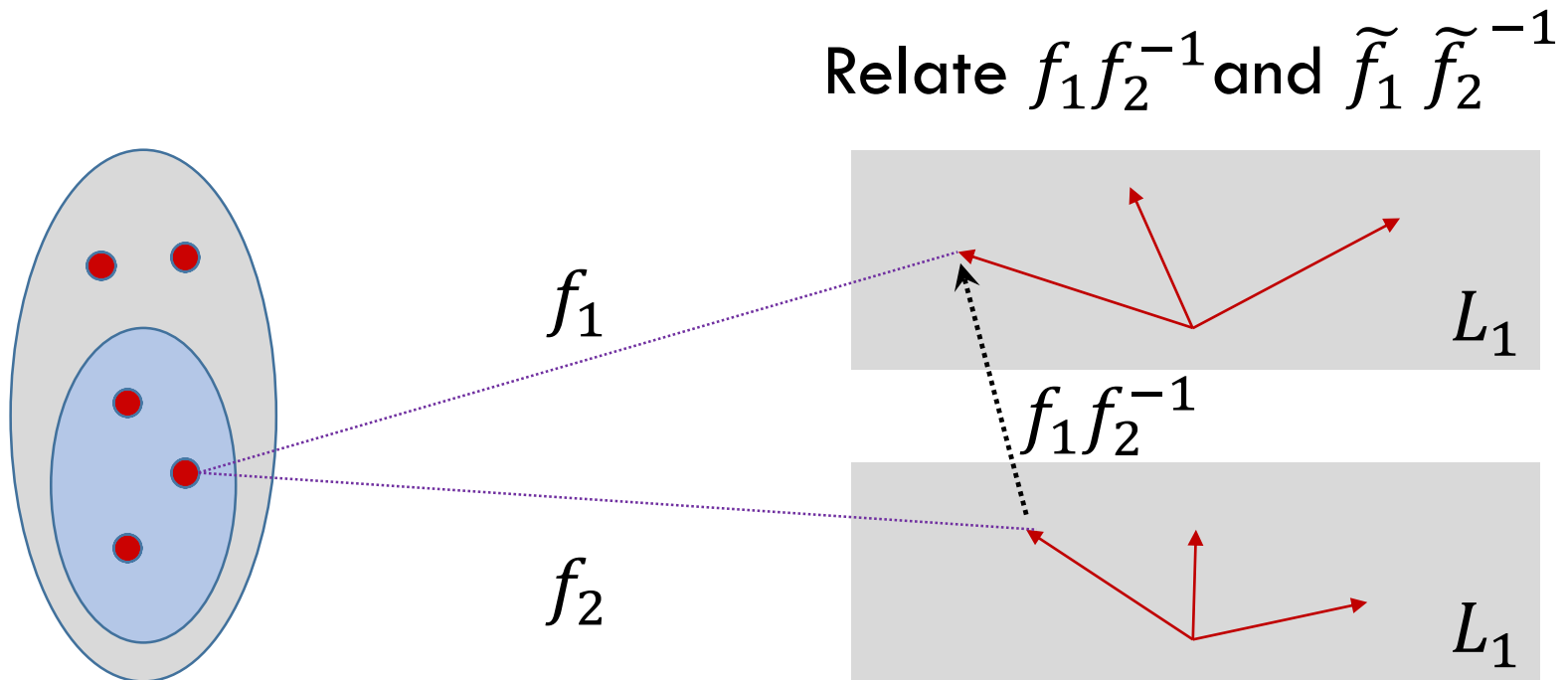
$$\mathbb{E}[\text{bk}_G(S)] = \min_{\tilde{f}} \sum_{u,v \in V} c'(u, v) \cdot \|\tilde{f}(u) - \tilde{f}(v)\|_1$$

Bob: gives **maps** $f_1: T \rightarrow L_1$ and $f_2: T \rightarrow L_1$

Need:

$$\text{bk}_G(\tilde{f}_1) \leq \text{bk}_H(f_1)$$

$$\text{bk}_H(f_2) \leq Q \text{bk}_G(\tilde{f}_2)$$

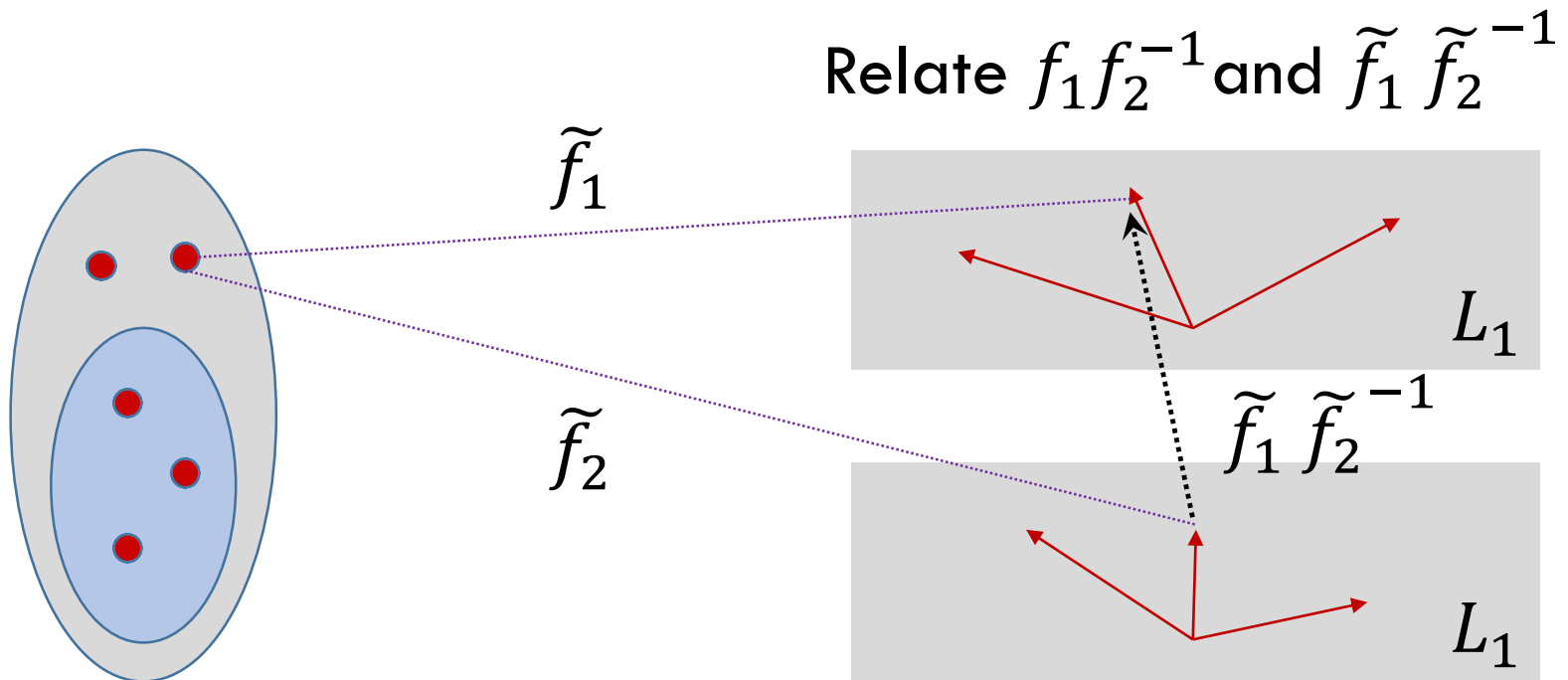


Bob: gives **maps** $f_1: T \rightarrow L_1$ and $f_2: T \rightarrow L_1$

Need:

$$\text{bk}_G(\tilde{f}_1) \leq \text{bk}_H(f_1)$$

$$\text{bk}_H(f_2) \leq Q \text{bk}_G(\tilde{f}_2)$$



Ball's Open Problem & Sparsification

[MM '10]

$$Q_k^{cut} = e_k(\ell_1, \ell_1) \leq e_k(\ell_2, \ell_1) \cdot C \sqrt{\log k} \log \log k$$

here, $C \sqrt{\log k} \log \log k$ is the distortion of the Fréchet embedding of ℓ_1 into ℓ_2 by Arora, Lee, Naor '07.

If $e_k(\ell_2, \ell_1) \leq C_{\text{Ball}}$ then

$$Q_k^{cut} \leq C' \sqrt{\log k} \log \log k$$

Outer bi-Lipschitz extendibility

Bi-Lipschitz Kirszbraun Theorem?

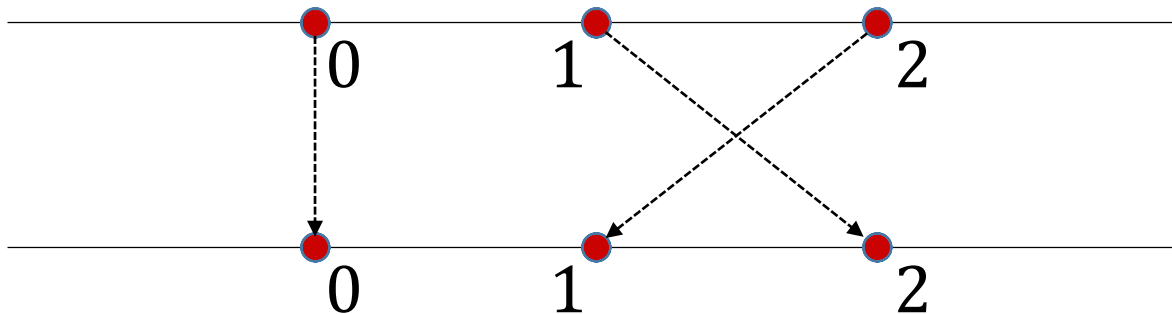
Let $A \subset \ell_2^m$. Can we extend a bi-Lipschitz map $f: A \rightarrow \ell_2^n$ to a bi-Lipschitz map $\tilde{f}: \ell_2^m \rightarrow \ell_2^n$?

No!

- $\mathbb{R}^2 \rightarrow \mathbb{R}$. Assume $\mathbb{R} \subset \mathbb{R}^2$. Extend $f = id_{\mathbb{R}}$?

There is even no **injective extension** of f to \mathbb{R}^2 .

- $\mathbb{R} \rightarrow \mathbb{R}$. f maps 0, 1, 2 to 0, 2, 1, respectively. There is even no **continuous one-to-one extension**.



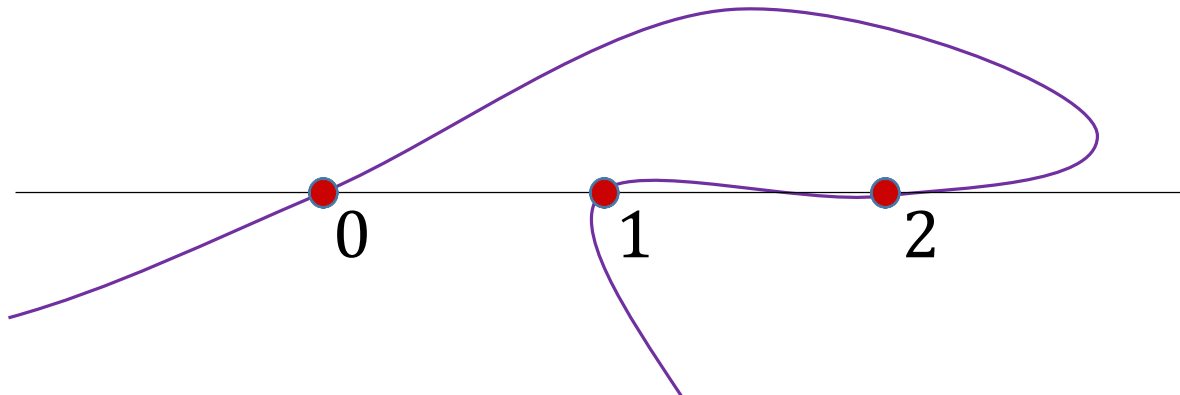
Bi-Lipschitz Kirszbraun Theorem?

Can overcome these obstacles by using extra dimensions!

• $\mathbb{R}^2 \rightarrow \mathbb{R}$. Assume $\mathbb{R} \subset \mathbb{R}^2$. Extend $f = id_{\mathbb{R}}$?

Assume that target $\mathbb{R} \subset \mathbb{R}^2$. Then $\tilde{f} = id_{\mathbb{R}^2 \rightarrow \mathbb{R}^2}$

• $\mathbb{R} \rightarrow \mathbb{R}$. f maps 0, 1, 2 to 0, 2, 1, respectively.



Outer bi-Lipschitz extension

Given $A \subset \ell_2^m$ and bi-Lipschitz $f: A \rightarrow \ell_2^n$.

Assume $\ell_2^n \subset \ell_2^N$.

$\tilde{f}: \ell_2^m \rightarrow \ell_2^N$ is an *outer bi-Lipschitz extension* of f if

$$\tilde{f}(a) = f(a) \text{ for every } a \in A$$

and \tilde{f} is bi-Lipschitz.

[MMMR '18] For every bi-Lipschitz $f: A \rightarrow \ell_2^n$, there exists an outer bi-Lipschitz extension with

$$D(\tilde{f}) \leq 3D(f).$$

Outer bi-Lipschitz extension

Near isometric case. What happens when

$$D(f) = 1 + \varepsilon ?$$

If $\varepsilon = 0$, i.e., f is an isometric map, there is an isometric extension \tilde{f}

Is there a bi-Lipschitz extension with

$$D(\tilde{f}) = 1 + o(1) \text{ as } \varepsilon \rightarrow 0 ?$$

For $f: A \rightarrow \mathbb{R}, A \subset \mathbb{R}$, **yes!** There is an extension with

$$D(\tilde{f}) = 1 + \frac{1}{\log^2 1/\varepsilon}$$

The bound is tight.

Outer bi-Lipschitz extension

Near isometric case. What happens when

$$D(f) = 1 + \varepsilon ?$$

If $\varepsilon = 0$, i.e., f is an isometric map, there is an isometric extension \tilde{f}

Is there a bi-Lipschitz extension with

$$D(\tilde{f}) = 1 + o(1) \text{ as } \varepsilon \rightarrow 0 ?$$

For a one-point extension of $f: A \rightarrow \ell_2^n$

$$D(\tilde{f}) = 1 + C\sqrt{\varepsilon}$$

The bound is tight.

Summary

Characterized the optimal distortion of cut and flow vertex sparsifiers in terms of Lipschitz extension constants.

- Find $e_k(\ell_1, \ell_1)$ and $e_k(\ell_\infty, \ell_\infty \oplus_1 \dots \oplus_1 \ell_\infty)$
- Is $e_k(\ell_2, \ell_1) < \infty$?

Defined outer bi-Lipschitz extension and proved an analogue of Kirzbraun theorem for it. Partial results for nearly isometric maps.

- Understand the nearly isometric case.

Applications to a prioritized dimension reduction.