# Two-sided Kirszbraun Theorem 

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## Plan

1. Background

- Extension of functions

2. Our results

- Two-sided Kirszbraun Theorem

3. Overview of the approach

## Extension of Functions

Notation throughout the talk

- We have a function $f: S \rightarrow \mathbb{R}^{n}$
- Which is defined over a subset $S \subset \mathbb{R}^{m}$


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$\square$ Extension to the superset $T$, i.e., $f^{\prime}: T \rightarrow \mathbb{R}^{m}$ so that

- $\boldsymbol{f}^{\prime}(\boldsymbol{x})=\boldsymbol{f}(\boldsymbol{x})$ for any $x \in S$
- Maintaining other properties ..


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(Lipschitz Extension)
2. Bi-Lipschitz Constant, i.e., distortion
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## Lipschitz Extension

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Kirszbraun extension theorem '34: for $\boldsymbol{S} \subset \mathbb{R}^{\boldsymbol{n}}$, every L-Lipschitz map $\boldsymbol{f}: \boldsymbol{S} \rightarrow \mathbb{R}^{\boldsymbol{m}}$ can be extended to the whole $\mathbb{R}^{n}$ keeping the same Lipschitz constant, i.e., $L^{\prime}=L$.

## Kirszbraun Extension Theorem

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\left\|f(x)-f\left(x^{\prime}\right)\right\| \leq L \cdot \| x \square \square \text { Applications }
$$

$\square$ Lipschitz extension:
Given: a L-Lipschitz map $f: S$ Goal: a $\operatorname{map} f^{\prime}: \mathbb{R}^{\boldsymbol{n}} \rightarrow \mathbb{R}^{\boldsymbol{m}}$ s.t.

- $f^{\prime}$ is an extension of $f$
- $f^{\prime}$ is $L^{\prime}$-Lipschitz
- Prioritized and Terminal Dimension reduction
- Clustering
- ...

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- Bi-Lipschitz extension: [MMMR'18]
- Initial map $f: X \rightarrow Y$ is $D$-bi-Lipschitz or has distortion $D$, i.e., for some $\lambda$ and all $x, x^{\prime} \in X$ :

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\lambda \cdot\left\|x-x^{\prime}\right\| \leq\left\|f(x)-\boldsymbol{f}\left(x^{\prime}\right)\right\| \leq D \cdot \lambda \cdot\left\|x-x^{\prime}\right\|
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- $f^{\prime}$ will have distortion $O(D)$
- Using "extra coordinates"


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\lambda \cdot\left\|\boldsymbol{x}-\boldsymbol{x}^{\prime}\right\| \leq\left\|\boldsymbol{f}(\boldsymbol{x})-\boldsymbol{f}\left(\boldsymbol{x}^{\prime}\right)\right\| \leq D \cdot \lambda \cdot\left\|\boldsymbol{x}-\boldsymbol{x}^{\prime}\right\|
$$

- $f^{\prime}$ will have distortion $O(D)$
- Using "extra coordinates"
$>$ What if we have no such guarantee?


## Two-Sided Kirszbraun Theorem

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- $x, y \in S$
- $\|f(x)-f(y)\|$ is small
- $x^{\prime} \in T$ is close to $x$, and $y^{\prime} \in T$ is close to $y$

$>$ Question: Can we decrease distances between any pair of points as little as possible?

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## Results in a nutshell

## A "tight" variant of the Kirszbraun theorem:

It is possible to find an extension map $f^{\prime}$ such that the distance between any pair of points is not decreased by more than what is "necessary".

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$$
\left(L\left\|x^{\prime}-x\right\|+\|f(x)-f(y)\|+L\left\|y-y^{\prime}\right\|\right)
$$



## What is necessary? An upper bound

$$
\inf _{x, y \in S}\left(L\left\|x^{\prime}-x\right\|+\|f(x)-f(y)\|+L\left\|y-y^{\prime}\right\|\right)
$$



## What is necessary? An upper bound

$$
\min \left(L\left\|x^{\prime}-y^{\prime}\right\| \inf _{x, y \in S}\left(L\left\|x^{\prime}-x\right\|+\|f(x)-f(y)\|+L\left\|y-y^{\prime}\right\|\right)\right.
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## What is necessary? An upper bound

- Define metric

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- cL-Lipschitz extension $f^{\prime}: T \rightarrow \mathbb{R}^{m}$ has $\left\|f^{\prime}\left(x^{\prime}\right)-f^{\prime}\left(y^{\prime}\right)\right\| \leq c d_{u b}\left(x^{\prime}, y^{\prime}\right)$



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- Can we find $\boldsymbol{f}^{\prime}$ such that $\left\|f^{\prime}\left(x^{\prime}\right)-f^{\prime}\left(y^{\prime}\right)\right\| \geq \Omega\left(d_{u b}\left(x^{\prime}, y^{\prime}\right)\right)$ ?



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- Short Answer: No
- Long Answer: We need extra relaxations


A bad example

- $S=C \cup\{x, y\}$



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- $f$ is 1-Lipschitz




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- $S=C \cup\{x, y\}, T=C \cup[x, y]$
- $f$ is 1-Lipschitz
- $d_{u b}>0$ for every pair of points $u, v \in T$
- But the image of $[x, y]$ intersects the circle, i.e., $f^{\prime}(u)-f^{\prime}(v)=0$




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## Relaxation I: Outer Extension

- Use additional coordinates in the image of the extended map




## Lipschitz Outer-Extension

Given: a map $f: S \rightarrow \mathbb{R}^{m}$, where

- $S \subseteq T \subset \mathbb{R}^{\boldsymbol{n}}$
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Goal: $\operatorname{a~map} f^{\prime}: X \rightarrow \mathbb{R}^{m^{\prime}}$, where

- $m^{\prime}>m$
- $f^{\prime}$ is $L^{\prime}$-Lipschitz

- $f^{\prime}$ is an (outer)-extension of $f$ : for every $x \in S$

$$
f^{\prime}(x)=f(x) \bigoplus(0, \ldots, 0)
$$

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- If $f^{\prime}$ must have Lipschitz constant equal to 1 , then $m$ should be mapped to $m^{\prime}$
- the distances would decrease infinitely.
$>$ Instead find a $(\mathbb{1}+\epsilon) L$-extension $f^{\prime}$




## Results: Two-sided Kirszbraun Theorem

## Given:

- $f: S \rightarrow \mathbb{R}^{m}$ is $L$-Lipschitz
- $S \subset T \subset \mathbb{R}^{n}$

Find: the extended map $f^{\prime}: T \rightarrow \mathbb{R}^{m} \oplus \mathbb{R}^{\Delta} \approx \mathbb{R}^{m^{\prime}}$ such that

- $f^{\prime}$ is $(1+\epsilon) L$-Lipschitz
- $\left\|f^{\prime}(x)-f^{\prime}(y)\right\| \geq c \sqrt{\epsilon} d_{u b}(x, y)$ for all $x, y \in T$
- If $|T \backslash S|$ is finite, then $\Delta=O(\log |T \backslash S|)$.
- Otherwise $\Delta=\infty$


## Pros

$>$ Least Possible contraction: for any pair simultaneously upto a factor of $O(\sqrt{\epsilon})$, i.e., Bound $\frac{\left\|f^{\prime}(x)-f^{\prime}(y)\right\|}{d_{u b}(x, y)} \in[c \sqrt{\epsilon}, 1+\epsilon]$

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$>$ Easy to compute distances $\left\|f^{\prime}(x)-f^{\prime}(y)\right\|$

- Computing $\left\|f^{\prime}(x)-f^{\prime}(y)\right\|$ in Kirszbraun theorem requires computing the entire map itself, which can be done using SDP
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> Optimal Parameters (See next slide)


## Lower bound results

1. $\sqrt{\epsilon}$ loss is required: There exists $S$ and $T=S \cup\left\{Z_{1}, Z_{2}\right\}$ and a 1-Lipschitz function $f$ s.t. for any $(1+\epsilon)$-Lipschitz extension of $f$, their distance has to decrease by a factor of $\sqrt{\epsilon}$, i.e., $\left\|f^{\prime}\left(z_{1}\right)-f^{\prime}\left(z_{2}\right)\right\| \leq O\left(\sqrt{\epsilon} d_{u b}\left(z_{1}, z_{2}\right)\right)$

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2. $\log |T \backslash S|$ dimensions is required for finite sets: for any $m, n, N$, there exists an instance s.t. $|T \backslash S|=N$, and any outer Lipschitz extension with \|f $f^{\prime}(x)-$ $f^{\prime}(y) \| \geq \boldsymbol{c} d_{u b}(x, y)$ requires $m^{\prime}=c^{\prime} \log N$ where $c^{\prime}=1 / \log \left(\frac{L}{c}+1\right)$

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3. Infinite dimension is required for infinite sets: for any $m, n$, there exists an instance with infinite sets $S \subset T$, s.t. any outer Lipschitz extension with $\left\|f^{\prime}(x)-f^{\prime}(y)\right\| \geq$ $c d_{u b}(x, y)$ for some $c$, requires $m^{\prime}=\infty$

## Application I: Bi-Lipschitz extension

- Our results immediately implies bi-Lipschitz extension of [MMakarychevMakarychevRazenshteyn'18]
- $O(D)$ distortion
- Caveat: we don't have $m^{\prime}=m+n$,
- Pros: easy to compute distances approximately.

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## Theorem:

- Sufficient conditions for it: if $d_{Y}(x, y) \leq C d_{X}(x, Y)$ for all $x, y \in Y$, then the updated metric is $\boldsymbol{O}(\boldsymbol{C A B})$-Euclidean, where we assume $d_{x}$ is $A$-Euclidean and $d_{y}$ is $B$-Euclidean


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- Lower bound: The above condition is necessary otherwise one gets at least $\Omega(\log N)$ distiortion


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2. $h^{\prime}(x)=c \sqrt{\epsilon} L h(x):$

- $h(x)$ should be 0 when $x \in S$
- Increases as a function of $R_{x}:=\operatorname{dist}(x, S)$
- $\|h(x)-h(y)\| \approx \Theta\left(\min \left(\|x-y\|, R_{x}+R_{y}\right)\right)$
- Use [Mendel\&Naor'04] embedding (rescaled and truncated)


## Construction of $h$

## Two ingredients:

## 1. [Mendel\&Naor'04]:

- For any $r>0$, there exists a map $\psi_{r}$ from $\ell_{2}^{n}$ to the infinite dimensional sphere of radius $r$, such that it approximately preserve distances of value at most $r$.
- $\|\psi(x)-\psi(y)\|=\Theta(\min (\|x-y\|, \sqrt{2} r))$


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Set $r \approx \boldsymbol{R}_{\boldsymbol{x}}$

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- $\|\psi(x)-\psi(y)\|=\Theta(\min (\|x-y\|, \sqrt{2} r))$


## 2. Bump Function:

$$
\lambda(t)= \begin{cases}e^{-\frac{1}{1-t^{2}}}, & \text { if } t \in(-1,1) \\ 0, & \text { otherwise }\end{cases}
$$



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1. $g^{\prime}(x)$ : Kirszbraun extension of $f(x)$.

- Does not increase distances
- The same as $f(x)$ on the points in $S$

2. $h^{\prime}(x)=c \sqrt{\epsilon} L h(x)$ :

- $h(x)$ should be 0 when $x \in S$
- Increases as a function of $R_{x}:=\operatorname{dist}(x, S)$
- $\|h(x)-h(y)\| \approx \Theta\left(\min \left(\|x-y\|, R_{x}+R_{y}\right)\right)$
- Use [Mendel\&Naor'04] embedding (rescaled and truncated)
- For finite set $|T \backslash S|$, we apply JL on top of $\boldsymbol{h}^{\prime}(\boldsymbol{x})$ to get the desired bound on the dimension


## Summary

- Showed two sided variant of the Kirszbraun theorem
- It achieves asymptotically optimal parameters.
- Provides a simple approximate formula for computing distances
- Applications of our results to bi-Lip extension \& Updating Euclidean metric.


## Given:

- $f: S \rightarrow \mathbb{R}^{m}$ is $L$-Lipschitz
- $\quad S \subset T \subset \mathbb{R}^{n}$

Find: the extended $\operatorname{map} f^{\prime}: T \rightarrow \mathbb{R}^{m} \oplus \mathbb{R}^{\Delta} \approx \mathbb{R}^{m^{\prime}}$ such that

- $f^{\prime}$ is $(1+\epsilon) L$-Lipschitz
- $\quad\left\|f^{\prime}(x)-f^{\prime}(y)\right\| \geq c \sqrt{\epsilon} d_{u b}(x, y)$ for all $x, y \in T$
- If $|T \backslash S|$ is finite, then $\Delta=O(\log |T \backslash S|)$.
- Otherwise $\Delta=\infty$


## Thanks!

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