Journée statistique et informatique pour la science des données IHES, January 2020

## The pre-image problem

from a topological perspective

## Steve Oudot

## Preimage problem in data Sciences



## Data



Features

bag of words, word2vec shape contexts, heat kernels node2vec, Laplacian fact., rand. walks dim. reduction, auto-encoders, etc.

## Preimage problem in data Sciences



## Data

## Features



Can the feature map be inverted?

- Right inverse ( $\exists$ preimage): interpretable AI bag of words, word2vec shape contexts, heat kernels
- Left inverse ( $\exists$ ! preimage): reliable interpretation

Scenarios: feature interpretation, deep learning, inverse problems, etc.

## Topological Data Analysis (TDA) pipeline



Data


Invariants


Features

## Topological Data Analysis (TDA) pipeline



## Invariants



Features

## Mathematical framework:



## Topological Data Analysis (TDA) pipeline




Features

## Mathematical framework:



## Topological Data Analysis (TDA) pipeline



## Mathematical framework:



## Topological Persistence (in a nutshell)

$X$ topological space $f: X \rightarrow \mathbb{R}$ (filter)
$\nabla$



signature: persistence diagram encodes the topological structure of the pair $(X, f)$


## Topological Persistence (in a nutshell)

Inside the black box:

- Nested family (filtration) of sublevel-sets $f^{-1}((-\infty, \alpha])$ for $\alpha$ ranging over $\mathbb{R}$
- Track the evolution of the topology throughout the family (homology)



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- Finite set of intervals (barcode) encodes births/deaths of topological features


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- Track the evolution of the topology throughout the family (homology)
- Finite set of intervals (barcode) encodes births/deaths of topological features
- Equivalent representation as a discrete measure in the plane (pers. diagram).



## Topological Persistence (in a nutshell)

## Theorem (Stability):

For any tame functions $f, g: X \rightarrow \mathbb{R}, \mathrm{~d}_{\mathrm{b}}^{\infty}(\operatorname{Dgm} f, \operatorname{Dgm} g) \leq\|f-g\|_{\infty}$.



## Example: distance function

$$
\begin{array}{ll}
f: & X=\mathbb{R}^{d} \rightarrow \mathbb{R} \\
& x \mapsto \min _{p \in P}\|x-p\|_{2}
\end{array}
$$



## Example: distance function

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## Vanilla pipeline



## Vanilla pipeline


metric space $(P, \mathrm{~d})$


Union of balls / Rips complex

$X=2^{P} \backslash\{\emptyset\}$
$f: \sigma=\left\{p_{0}, \cdots, p_{k}\right\} \mapsto \max _{1 \leq i<j \leq k} \mathrm{~d}\left(p_{i}, p_{j}\right)$
Rips filtration $\mathcal{R}(P): R_{t}(P):=f^{-1}((-\infty, t])$


## Many variants (filters, topological constructions, approximations)


density estimators


- non-linear projections
- curvature measures
- PDE solutions (heat, wave)
- etc.

projections


## others:

## Software

##  in Higher Dimensions

```
http://gudhi.inria.fr/
```

- reference library in TDA (encompasses all aspects)
- 60 k downloads in the last 12 months
- developers community, editorial board
- competitors (specialized on specific aspects of the TDA pipeline):

Dionysus, PHAT, DIPHA, Ripser, Eirene

## The problem

compact metric space
persistence barcode/diagram

right inverse: realize barcode as the PH of some isom. class
left inverse: characterize isometry class uniquely

## Right inverses for Topological Persistence

Fact: [Moore spaces] Any finitely generated Abelian group can be realized as the (reduced) homology of some topological space.

bouquets of spheres

## Right inverses for Topological Persistence

Fact: [Moore spaces] Any finitely generated Abelian group can be realized as the (reduced) homology of some topological space.

Thm: Any locally finite point cloud in $\mathbb{R}^{2}$ can be realized as the (extended) persistence diagram of some function on a topological space.

bouquets of spheres

handlebody theory

## Local right inverses



## Local right inverses


ordered point clouds with $n$ points in $\mathbb{R}^{d}$
ordered barcodes
with $m$ bounded and
$n$ unbounded intervals
( $\nexists$ smooth structure on Bar)

## Local right inverses



Thm: [Gameiro, Hiraoka, Obayashi '16]
(i) Generic point cloud $\Rightarrow \exists U \ni P$ in $\mathbb{R}^{n d}$ over which the mapping $P \mapsto v$ can be extended to a function $\tilde{B}: U \rightarrow \mathbb{R}^{2 m+n}$ computing ordered barcodes.
(ii) For $U$ small enough, $\tilde{B}$ is of class $C^{\infty}$.

Observation: order of distances is constant in small enough $U$.

## Local right inverses

Local lift:

$$
\begin{aligned}
& \exists \tilde{B} \quad \downarrow^{\mathbb{R}^{2 m+n}} \\
& \mathcal{M} \xrightarrow[F]{\longrightarrow \cdots \cdots \cdots} \operatorname{Filter}\left(K=2^{\{1, \cdots, n\} \backslash\{\emptyset\}}\right) \xrightarrow[\text { Dgm }]{ } \text { Bar } \\
& \text { || } \\
& \mathbb{R}^{n d}
\end{aligned}
$$

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## Local right inverses

$$
(\text { smooth }) \exists \tilde{B}
$$

Local lift:

$$
\mathcal{M} \xrightarrow[F]{\stackrel{\cdots \cdots \cdots}{\longrightarrow}} \operatorname{Filter}\left(K=2^{\{1, \cdots, n\} \backslash\{\emptyset\}}\right)
$$



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## Local right inverses

$$
\begin{equation*}
(\text { smooth }) \exists \tilde{B} \tag{Lip.}
\end{equation*}
$$

Local lift:

$$
\mathcal{M} \xrightarrow[F]{\stackrel{\cdots \cdots}{\longrightarrow}} \operatorname{Filter}\left(K=2^{\{1, \cdots, n\} \backslash\{\emptyset\}}\right)
$$



Diffeology theory: push smooth struct. on $\mathbb{R}^{2 m+n}$ down to Bar

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## Local right inverses



Thm: [Leygonie, O., Tillmann '19]
If $F$ is $C^{r}$ on some generic subset of $\mathcal{M}$, then so is $\operatorname{Dgm} \circ F$.

Prop: [Chain Rule]
If $\operatorname{Dgm} \circ F$ and $V$ are $r$-differentiable, then $V \circ \operatorname{Dgm} \circ F$ is $C^{r}$ and $d_{\theta}(V \circ \operatorname{Dgm} \circ F)=d_{\tilde{B}(\theta)}\left(V \circ Q_{m, n}\right) \circ d_{\theta} \tilde{B}$ is independent of lift

## Local right inverses

## Applications:

Optimization / ML: (gradient back-propagation)


## Local right inverses

## Applications:

Optimization / ML: (gradient back-propagation)


Continuation / inverse problems: (Newton-Raphson)



[Nakamura et al. '15]

## Left inverses?

- distance functions

$\operatorname{Dgm} \mathcal{R}(P)=\{(0,+\infty)\} \sqcup\{(0,1)\} \sqcup\{(0,1)\}$
$\Rightarrow$ diagrams for different values of $\alpha$ are indistinguishable


## Left inverses?

- distance functions

Prop: For any metric tree $(P, \mathrm{~d})$ :

$$
\operatorname{Dgm} \mathcal{R}(P)=\{(0,+\infty)\}
$$

$\Rightarrow$ no information on the metric
$X$ is 0-hyperbolic
$\Rightarrow$ metric balls are convex
$\Rightarrow$ geodesic triangles are tripods

## Left inverses?

- distance functions

Prop: For any metric tree $(P, \mathrm{~d})$ :

$$
\operatorname{Dgm} \mathcal{R}(P)=\{(0,+\infty)\}
$$

$\Rightarrow$ no information on the metric

- real-valued functions

Prop: For any $f: X \rightarrow \mathbb{R}$ and $h: Y \rightarrow X$ homeomorphism:

$$
\operatorname{Dgm} f \circ h=\operatorname{Dgm} f
$$

$\Rightarrow$ Invariance under homeomorphisms, not just isometries

## Persistent Homology Transform (PHT)



## PHT for compact subanalytic sets in $\mathbb{R}^{d}$

Focus: compact subanalytic sets in $\mathbb{R}^{d}$

PHT: $\mathcal{F}=\left\{f_{w}\right\}_{w \in \mathbb{S}^{d-1}}$ where $f_{w}=\langle\cdot, w\rangle$

Thm: [Boyer, Curry, Mukherjee, Turner 2014, 2018] [Ghrist, Levanger, Mai 2018]

With $\mathcal{F}=\{\langle\cdot, w\rangle\}_{w \in \mathbb{S}^{d-1}}$, PHT is injective on the class of compact subanalytic sets in $\mathbb{R}^{d}$.

Still true for a finite $\left(O\left(2^{d}\right)\right)$ set of directions if we restrict to geometric simplicial complexes.

## PHT for compact subanalytic sets in $\mathbb{R}^{d}$

Formalism:

$M \subset \mathbb{R}^{d}$<br>compact<br>subanalytic

$\leadsto$
$\mathbb{1}_{M}: \mathbb{R}^{d} \rightarrow \mathbb{Z}$
constructible

## PHT for compact subanalytic sets in $\mathbb{R}^{d}$


$\operatorname{ECT}\left(\mathbb{1}_{M}\right): \left\lvert\, \begin{aligned} & \mathbb{S}^{d-1} \rightarrow \mathrm{CF}(\mathbb{R}) \\ & v \longmapsto\left(t \mapsto \int \mathbb{1}_{M} \mathbb{1}_{x \cdot v \leq t} \mathrm{~d} \chi=\chi\left(M \cap\left\{x \in \mathbb{R}^{d} \mid x \cdot v \leq t\right\}\right)\right)\end{aligned}\right.$

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$$

Thm: For $\mathbb{1}_{M}, \mathbb{1}_{N} \in \operatorname{CF}\left(\mathbb{R}^{d}\right): \operatorname{ECT}\left(\mathbb{1}_{M}\right)=\mathrm{ECT}\left(\mathbb{1}_{N}\right) \Rightarrow \mathbb{1}_{M}=\mathbb{1}_{N}$

## PHT for compact subanalytic sets in $\mathbb{R}^{d}$

Formalism: $\begin{array}{ll} & M \subset \mathbb{R}^{d} \\ & \begin{array}{l}\text { compact } \\ \text { subanalytic }\end{array}\end{array}$


Thm: For $\mathbb{1}_{M}, \mathbb{1}_{N} \in \operatorname{CF}\left(\mathbb{R}^{d}\right): \operatorname{ECT}\left(\mathbb{1}_{M}\right)=\mathrm{ECT}\left(\mathbb{1}_{N}\right) \Rightarrow \mathbb{1}_{M}=\mathbb{1}_{N}$

NHWMy

## PHT for metric graphs

Focus: compact metric graphs (1-dimensional stratified length spaces)
PHT: $\mathcal{F}=\left\{\mathrm{d}_{X}(\cdot, x)\right\}_{x \in X}$

Thm (stability): [Dey, Shi, Wang 2015]
For any compact metric graphs $X, Y$,
$\mathrm{d}_{\mathrm{H}}(\operatorname{PHT}(X), \operatorname{PHT}(Y)) \leq 18 \mathrm{~d}_{\mathrm{GH}}(X, Y)$.

Thm (density): [Gromov]
Compact metric graphs are GH-dense among the compact length spaces.

Q: injectivity of PHT on metric graphs?

## PHT for metric graphs

Bad news: PHT is not injective on all compact metric graphs

$\operatorname{PHT}(X)=\operatorname{PHT}(Y)$ while $X \not 千 Y$

## PHT for metric graphs

Bad news: PHT is not injective on all compact metric graphs

$\operatorname{PHT}(X)=\operatorname{PHT}(Y)$ while $X \not 千 Y$

Note: $\operatorname{Aut}(X)$ is non-trivial, hence $\Psi_{X}: x \mapsto \operatorname{Dgmd}(\cdot, x)$ is not injective

## PHT for metric graphs

Let $\operatorname{Inj}_{\Psi}=\left\{X\right.$ compact metric graph s.t. $\Psi_{X}$ is injective $\}$

Thm: (injectivity) [O., Solomon '18]

- PHT is GH-locally injective on compact metric graphs.
- PHT is injective on $\operatorname{Inj}{ }_{\Psi}$.
- $\ln _{\Psi}{ }_{\Psi}$ is generic among the compact metric graphs.

PHT is injective on a dense subset of the compact length spaces

## PHT for compact metric measure spaces

Let $(X, \mathrm{~d}, \mu)$ be a compact metric (Borel) measure space
Distance kernel operator:

$$
\begin{aligned}
D^{X}: & L^{2}(X) \rightarrow L^{2}(X) \\
& \left(D^{X} f\right)(x):=\int_{X} f(y) \mathrm{d}(x, y) d \mu(y)
\end{aligned}
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Hilbert-Schmidt op. $\Rightarrow$ eigenvalues $\left|\lambda_{1}\right| \geq\left|\lambda_{2}\right| \geq \cdots$, assumed simple wlog
Choose unit eigenfunctions $\phi_{1}, \phi_{2}, \cdots$, such that $\left\langle\phi_{i},\right| \phi_{i}| \rangle>0$

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Mapping: $\Phi^{X}: X \rightarrow \mathbb{C}^{\infty}$

$$
\Phi^{X}:=\left(\sqrt{\lambda_{1}} \phi_{1}, \sqrt{\lambda_{2}} \phi_{2}, \cdots\right)
$$

$\Phi_{k}^{X}:=\Phi_{\mid \mathbb{C}^{k}}^{X}$
note:

$$
\begin{aligned}
\mathrm{d}(x, y) & =\sum_{i \in \mathbb{N}} \lambda_{i} \phi_{i}(x) \phi_{i}(y) \\
& =\sum_{i \in \mathbb{N}} \sqrt{\lambda_{i}} \phi_{i}(x) \sqrt{\lambda_{i}} \phi_{i}(y)
\end{aligned}
$$

## PHT for compact metric measure spaces

Let $(X, \mathrm{~d}, \mu)$ be a compact metric (Borel) measure space

Thm: [Maria, O., Solomon '19] If $\mu$ is strictly positive on open sets, then $\Phi^{X}: X \rightarrow \mathbb{C}^{\infty}$ is a topological embedding

Thm: [Maria, O., Solomon '19] Let $\mathrm{d}, \mathrm{d}^{\prime}$ be metrics on $X$, and let $\mu, \mu^{\prime}$ be strictly positive measures on $X$ such that $\mu$ is absolutely continuous w.r.t. $\mu^{\prime}$. Then,

$$
\Phi^{(X, \mathrm{~d}, \mu)}(X)=\Phi^{\left(X, \mathrm{~d}^{\prime}, \mu^{\prime}\right)}(X) \quad \Longrightarrow \quad \mathrm{d}=\mathrm{d}^{\prime}
$$

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$$

Pb: the Euclidean PHT and ECT apply only to finite-dimensional spaces

## PHT for compact metric measure spaces

Let $(X, \mathrm{~d}, \mu)$ be a compact metric (Borel) measure space
$\operatorname{Fix} k \in \mathbb{N}$.
$\Phi_{k}^{X}: X \rightarrow \mathbb{C}^{k} \simeq \mathbb{R}^{2 k}$ may not be an embedding... but it doesn't matter

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Let $(X, \mathrm{~d}, \mu)$ be a compact metric (Borel) measure space
$\operatorname{Fix} k \in \mathbb{N}$.
$\Phi_{k}^{X}: X \rightarrow \mathbb{C}^{k} \simeq \mathbb{R}^{2 k}$ may not be an embedding... but it doesn't matter
$X \mapsto \Phi^{X}$ may not be injective... but we can bound its fibers:

Thm: [Maria, O., Solomon '19] Assume $\Phi_{k}^{X}(X)=\Phi_{k}^{Y}(Y)$, under the same conditions as previously. Then, $\mathrm{d}_{\mathrm{GH}}(X, Y) \leq E_{X, k}+E_{Y, k}$, where $E_{X, k}$ measures the sup-norm difference between d and its order- $k$ eigenfunction expansion $(x, y) \mapsto \sum_{i=1}^{k} \lambda_{i} \phi_{i}(x) \phi_{i}(y)$.

Cor: [Maria, O., Solomon '19] Assume $\Phi_{k}^{X}(X)=\Phi_{k}^{Y}(Y)$, where $X, Y$ are finite or have finite non-zero spectrum. Then, under the same conditions as previously, and for $k$ large enough, $X$ and $Y$ are isometric.

Thank you

