



# Computer Aided Topological Design

## —Applications of computational topology in geometric design and processing

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- 1. Introduction to computational topology**
- 2. Applications in geometric design**
- 3. Porous retrieval, design, and printing**
- 4. Interpretability of DNN**
- 5. Applications in computational psychology**
- 6. Conclusion**

# Brief History

- Computational Topology is first presented in:  
Dey T K, Edelsbrunner H, Guha S. Computational topology. Contemporary mathematics, 1999, 223: 109-144.
- Computational Geometry → Computational Topology
- Topological Data Analysis: Proposed in 2009  
Carlsson G. Topology and Data. Bulletin of the American Mathematical Society, 2009, 46(2): 255-308.
- Computer Aided Geometric Design → Computer Aided Topological Design

1.2 创始人介绍

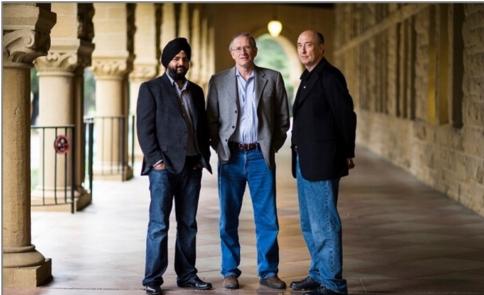
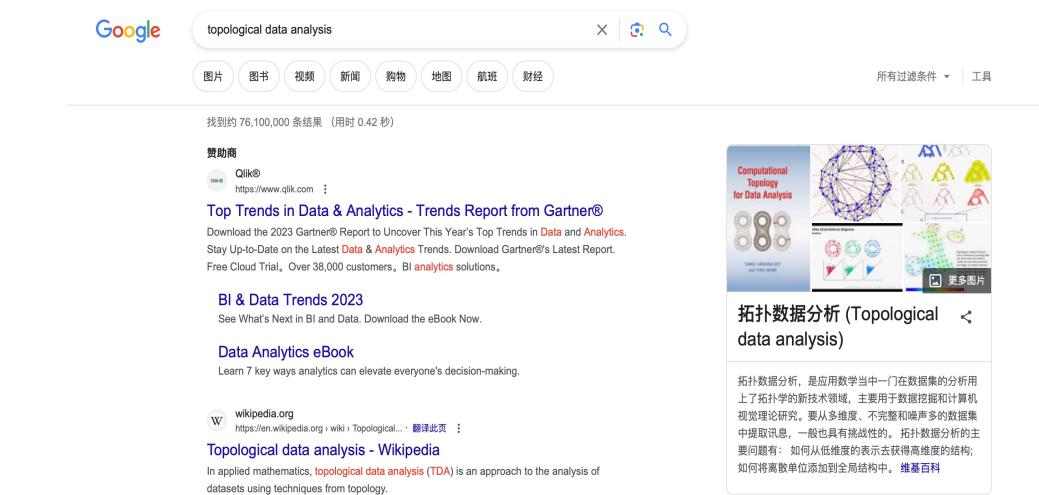


图: Ayasdi创始人Gurjeet Singh、Gunnar Carlsson、Harlan Sexton (从左往右)

2008年, Gunnar Carlsson、Gurjeet Singh和Harlan Sexton在斯坦福大学进行了12年的研发工作后联合创立了Ayasdi公司。在斯坦福大学研究期间,三位创始人获得了美国国防部高级研究计划局(DARPA)和美国情报高级研究计划署(IARPA)的资助,用于“高风险、高回报”的研究项目。2012年, Ayasdi获得了由Floodgate Capital和Khosla Ventures牵头的1025万美元的首轮融资,就此这家高精技术公司开启了它的征程。

公司总顾问Gunnar Carlsson是世界上最著名的数学家之一,他拥有哈佛大学的本科学位和斯坦福大学的博士学位,在过去的35年里,先后在芝加哥大学、加州大学、普林斯顿大学任教。自1991年以来,Gunnar一直在斯坦福大学担任数学教授,他是斯坦福大学数学领域里拓扑学研究的思想领袖。在21世纪初,这项工作获得了美国国家科学基金会(NSF)和美国国防部高级研究计划局(DARPA)1000万美元的研究拨款,用于研究拓扑数据分析技术(TDA)在美国政府关切问题上的应用。

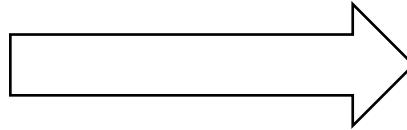
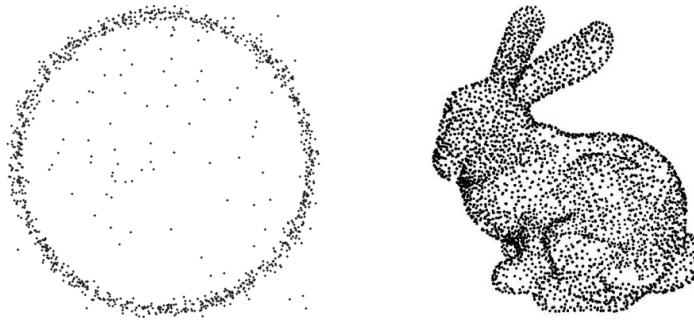
<https://www.weiyangx.com/379809.html>



The screenshot shows a Google search results page for the query "topological data analysis". The search bar at the top has the query entered. Below the search bar, there are several navigation links: 图片 (Images), 图书 (Books), 视频 (Videos), 新闻 (News), 购物 (Shopping), 地图 (Maps), 航班 (Flights), and 财经 (Finance). The main search results area displays a list of links, each with a snippet of text and a small thumbnail image. One prominent result is from Qlik, titled "Top Trends in Data & Analytics - Trends Report from Gartner®". Another result is from wikipedia.org, titled "Topological data analysis - Wikipedia". On the right side of the search results, there is a sidebar with the heading "拓扑数据分析 (Topological data analysis)". This sidebar contains a brief introduction to the topic, mentioning it is a branch of applied mathematics, and provides a link to the "维基百科" (Wikipedia) entry.

TDA已经成功应用于金融、生物、医学、药物设计、人工智能、脑神经科学等领域

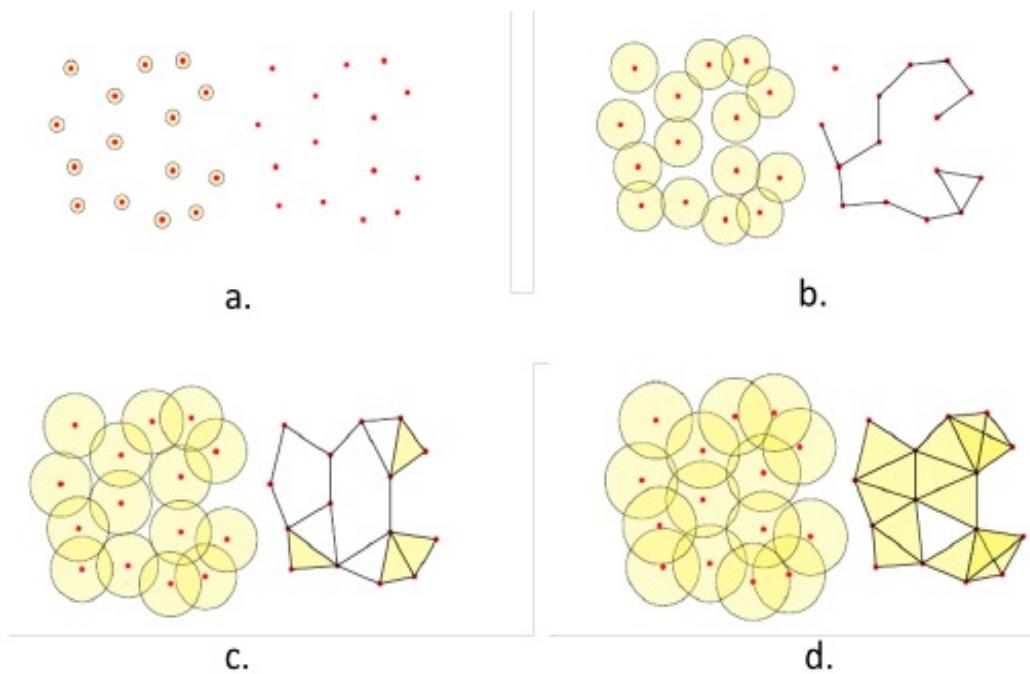
# Motivation of Topological Data Analysis



Topological Inference for  
point cloud data

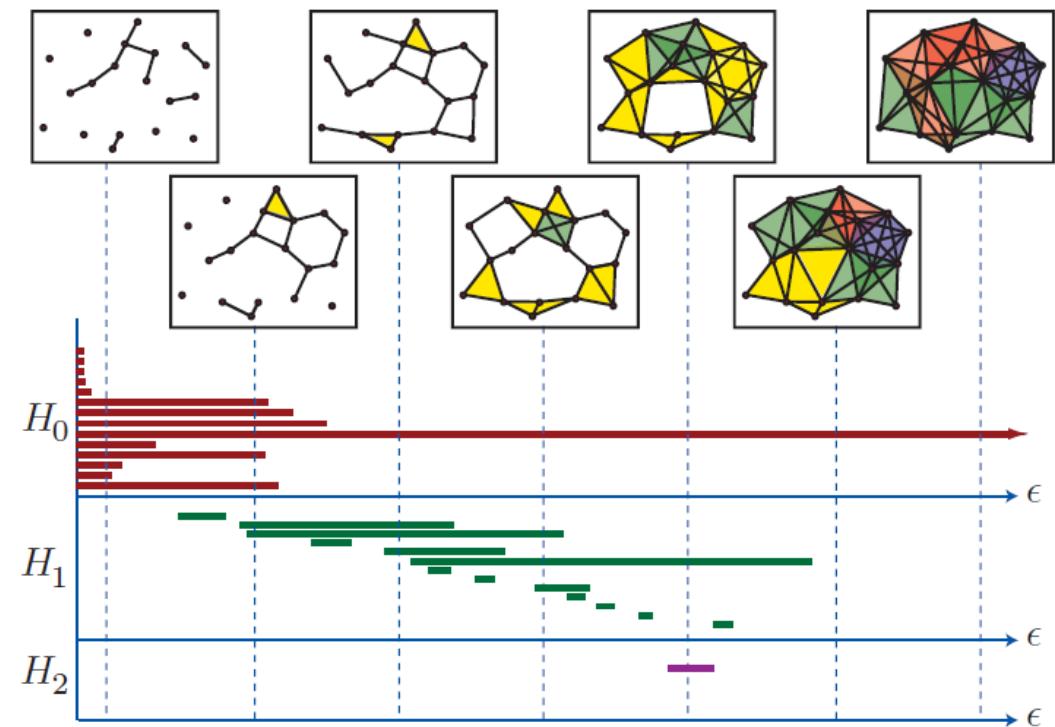
- How to **infer** the **essential topological features** of a hidden shape?
  - To extract components, loops, voids and higher dimensional features
  - To measure their importance
  - Stable with respect of small perturbations
- Mathematical Background
  - Algebraic Topology (Homology Theory)
  - Statistics
  - Geometry
- A promising bridge between topology and geometry in the view of computation

# Filtration, Persistent Homology



Filtration (Čech complex)

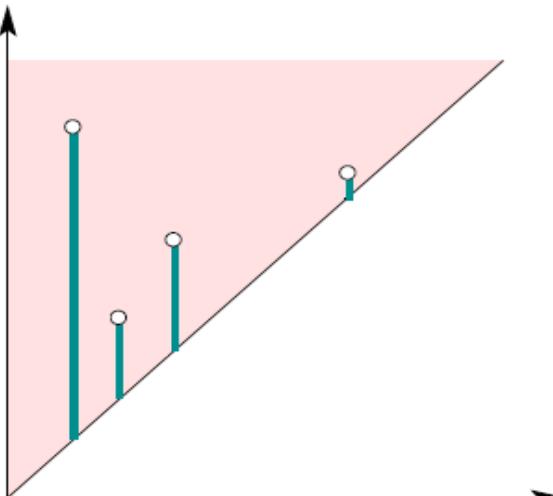
As the radii of open balls grow, the **simplicial complex** can be constructed on point cloud data, which gives a nested complex sequence, a **filtration**.



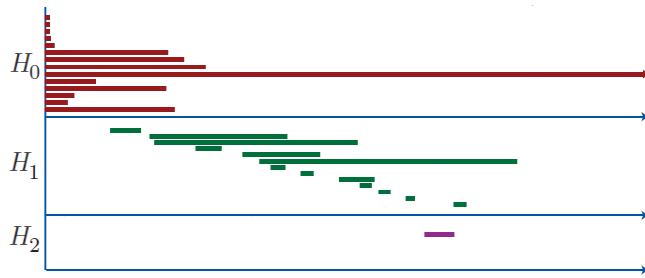
## Barcode

- As time goes by, the radii of open balls grow
- New  $k$ -holes are **born**
- Some  $k$ -holes are destroyed by higher-dimensional simplex and **die**

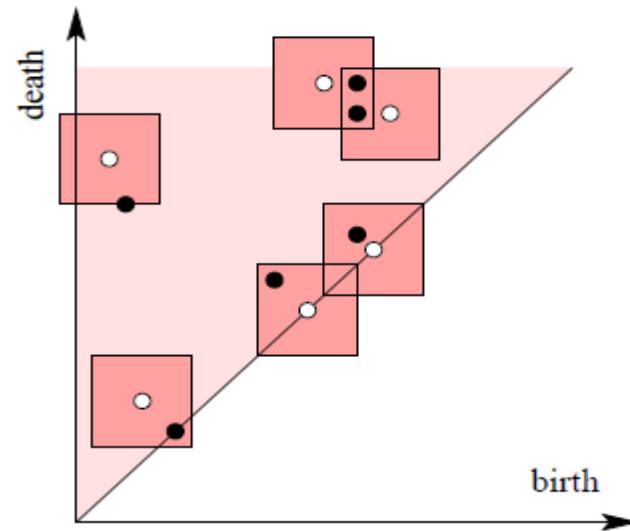
# Persistence Diagram, Distance and Stability



Persistence Diagram



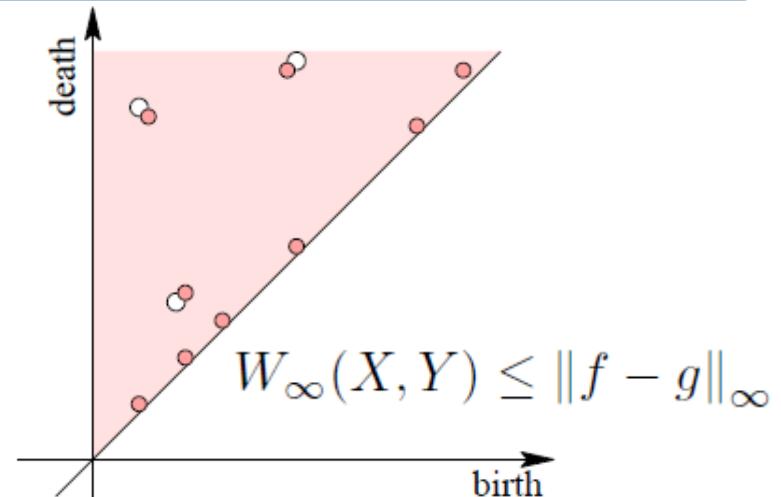
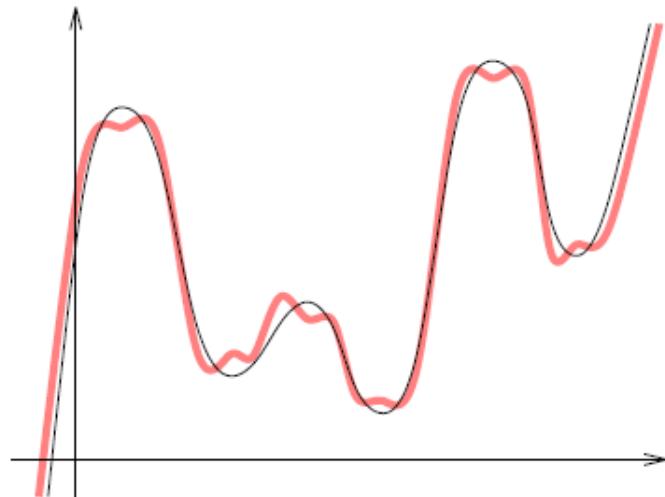
Barcode



$$W_\infty(X, Y) = \inf_{\eta: X \rightarrow Y} \sup_{x \in X} \|x - \eta(x)\|_\infty$$

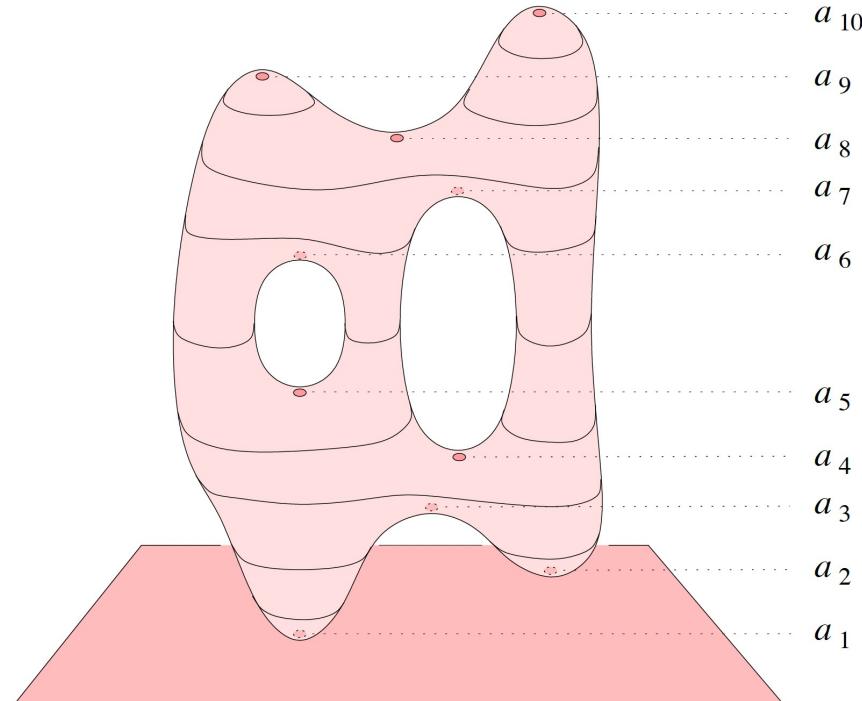
Wasserstein Distance

**Stability:** The *Wasserstein distance* of two persistence diagrams is controlled by the small perturbations on the tame function  $f$



$$W_\infty(X, Y) \leq \|f - g\|_\infty$$

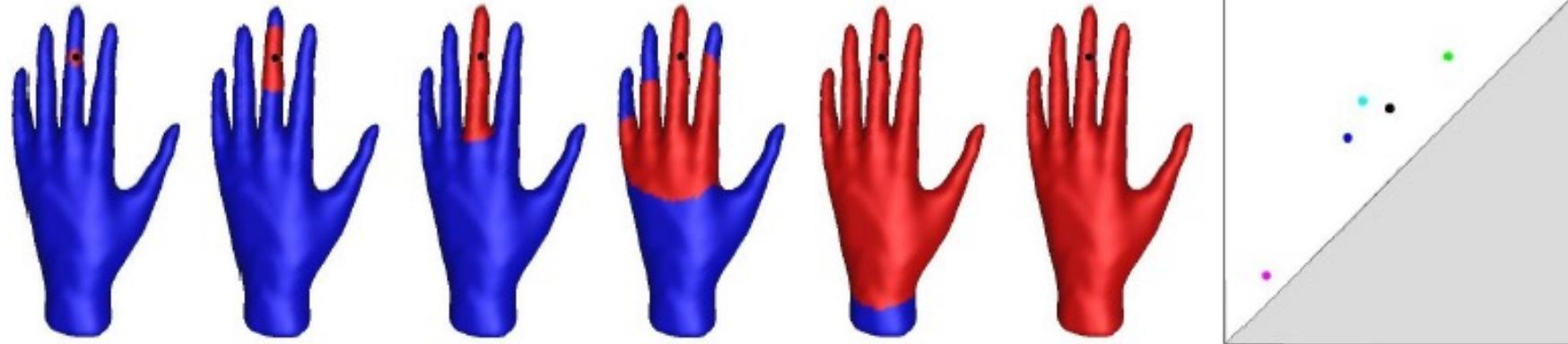
- Distance definition and computation
- Stability
- Representation
- Extended persistence
- Zigzag persistence
- Multiparameter persistence
- Persistence Module
- ...





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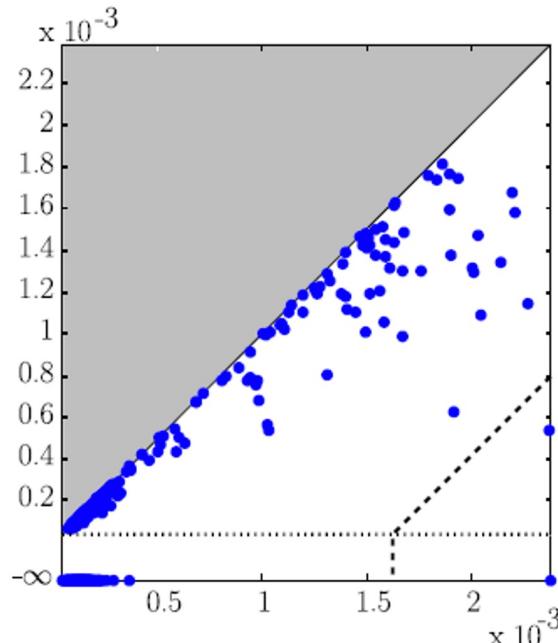
# Applications: Shape Description



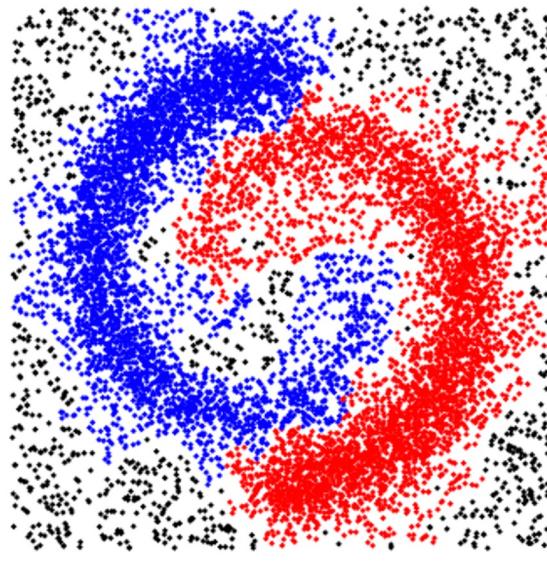
- The **persistence diagram** corresponding to this family is shown in the right
- The pink, blue, light blue, black and green points correspond to the middle, index, ring, pinky and thumb respectively

Carrière M, Oudot S Y, Ovsjanikov M. Stable topological signatures for points on 3d shapes. Computer graphics forum. 2015, 34(5): 1-12.

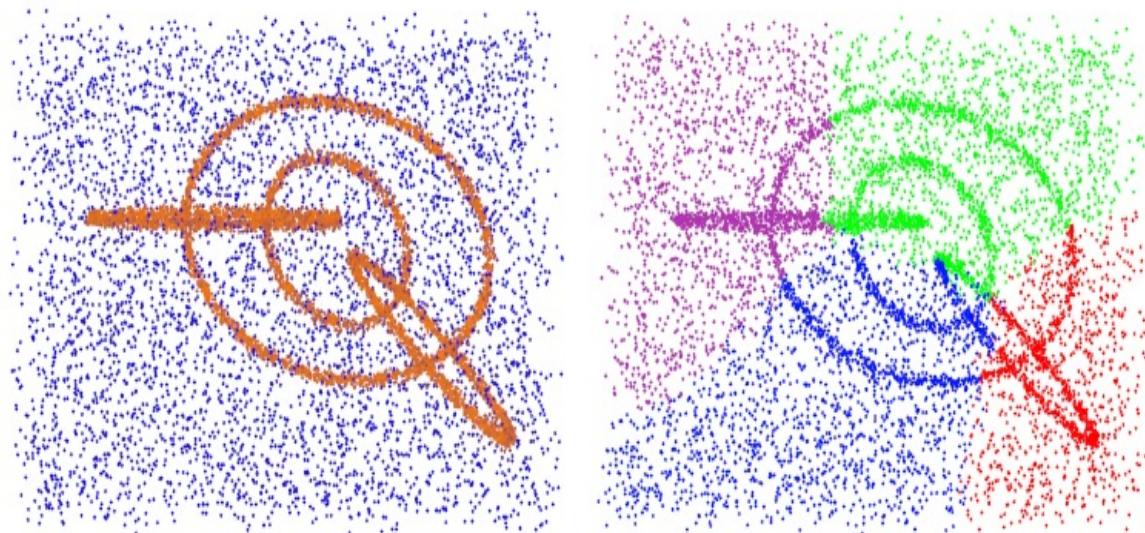
# Applications: Clustering of Point Clouds



(a)



(b)



- The rings are detected by the clustering method with **persistence diagram** (left)
- Compared with the result obtained by spectral clustering (right)

Chazal F, Guibas L J, Oudot S Y, et al. Persistence-based clustering in Riemannian manifolds[J]. Journal of the ACM (JACM), 2013, 60(6): 1-38.

# Applications: Topological Denoising

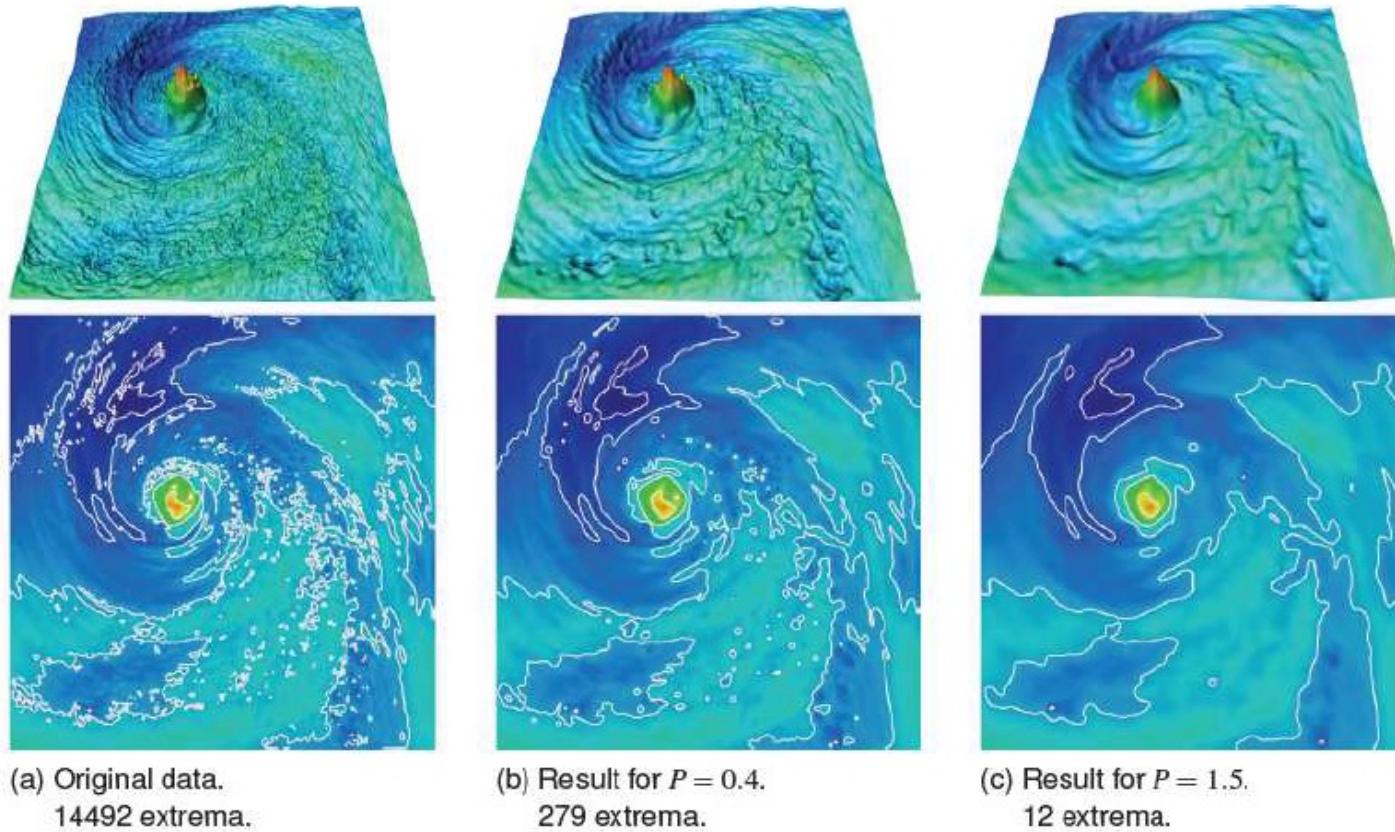


Figure 10. Temperature in the Hurricane Isabel data set (slice  $z = 20$ ). Using persistence-based filtering, we create a hierarchy of scalar fields: with increasing persistence  $P$ , our method creates increasingly smoother versions of the data.

- Keep the salient features while denoising by **persistence-based filtering**

Günther D, Jacobson A, Reininghaus J, et al. Fast and memory-efficient topological denoising of 2D and 3D scalar fields. IEEE transactions on visualization and computer graphics, 2014, 20(12): 2585-2594.

# Applications: Topology-aware Reconstruction

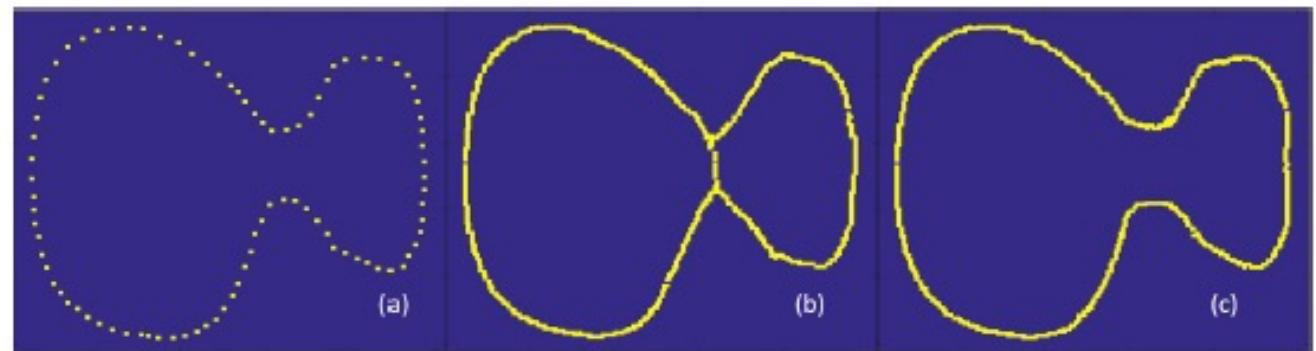
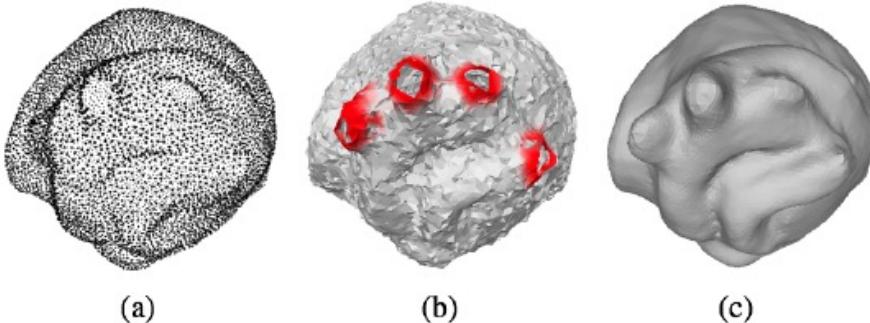
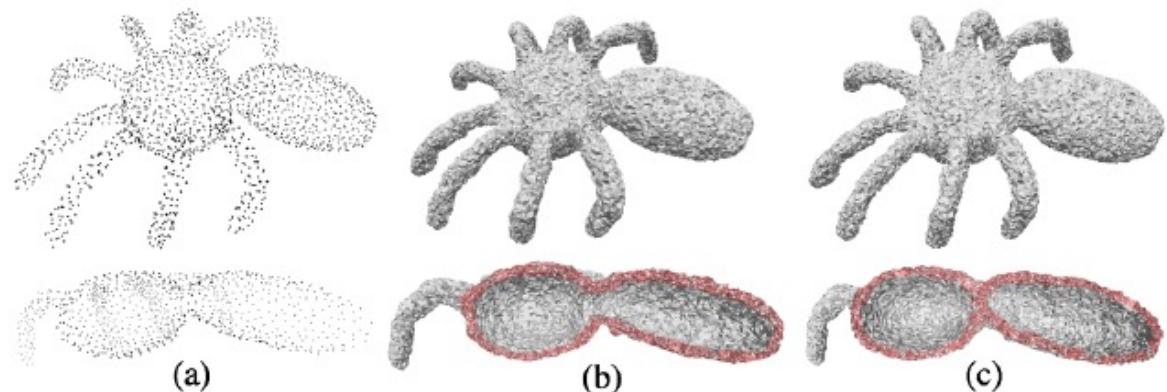
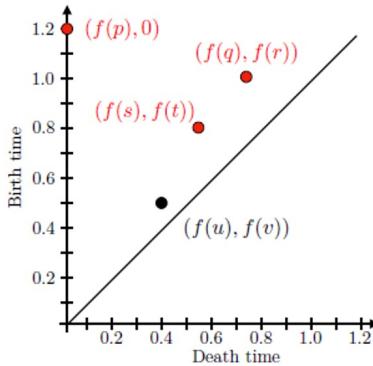
- **Objective Function:**

$$-((d_1 - b_1)^2 - (d_2 - b_2)^2)$$

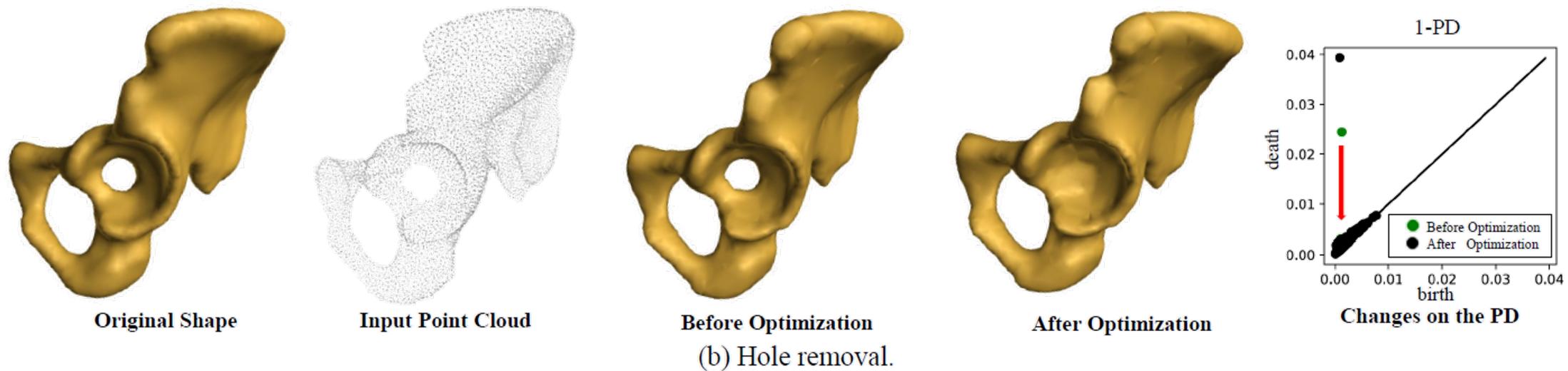
- The persistence of the first point is getting longer and longer
- The persistence of the other points are getting shorter and shorter.

$$-((d_2 - b_2)^2 - (d_3 - b_3)^2)$$

- The persistence of the first two point is getting longer and longer
- The persistence of the other points are getting shorter and shorter



Brüel-Gabrielsson, Rickard, et al. Topology-Aware Surface Reconstruction for Point Clouds. *Computer Graphics Forum*. Vol. 39. No. 5. 2020.



## Application: Hole removal

**Motivation:** To remove the 1D holes (1D homology classes) on the shape

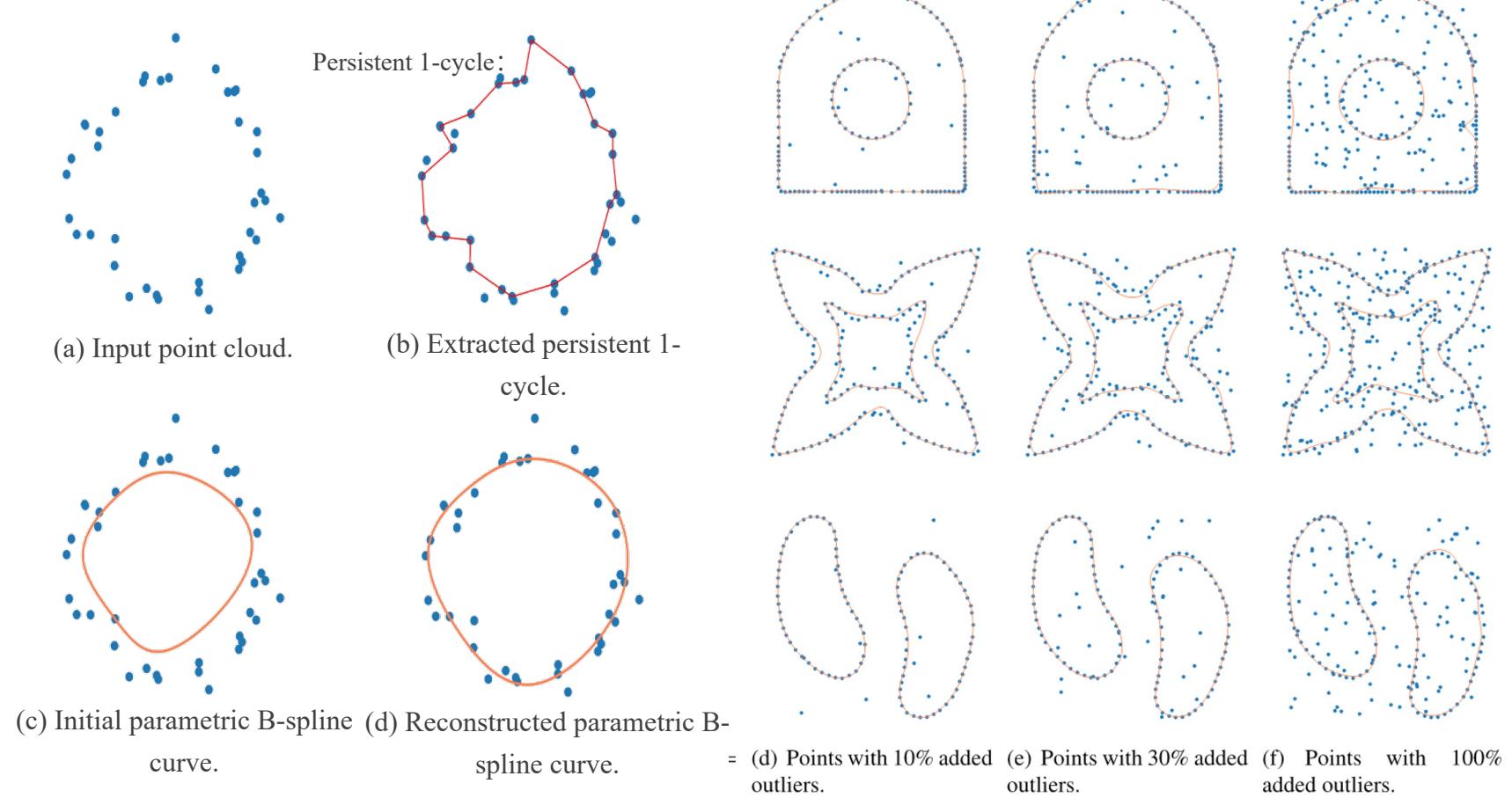
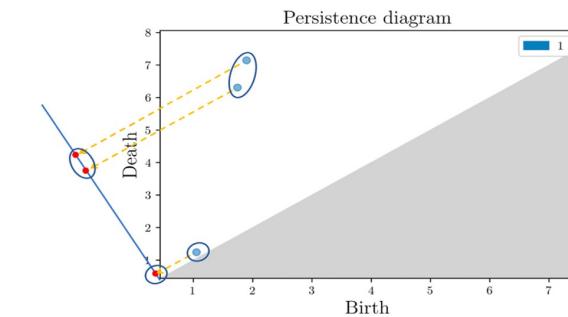
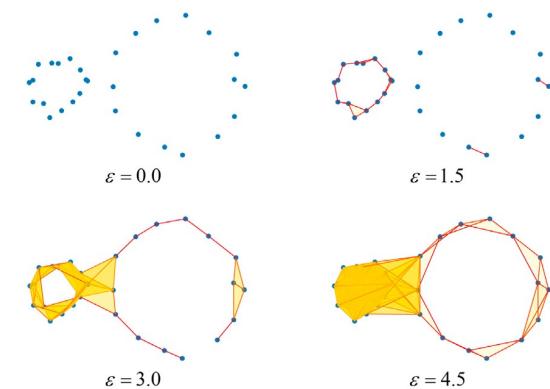
**Topology-controllable optimization:** To minimize  $L^{(1)} = d_2 - b_2$ , where  $(b_2, d_2)$  represents the persistence pair of the second most persistent feature

**Result:** The extracted surface was obtained with the removed small-scale handle loop

Zhetong Dong, Jinhao Chen, Hongwei Lin. Topology-controllable Implicit Surface Reconstruction Based on Persistent Homology. Computer-Aided Design, 150: 103308, 2022 (SPM 2022)

# Applications: Topology understanding and curve reconstruction

## Topology understanding:



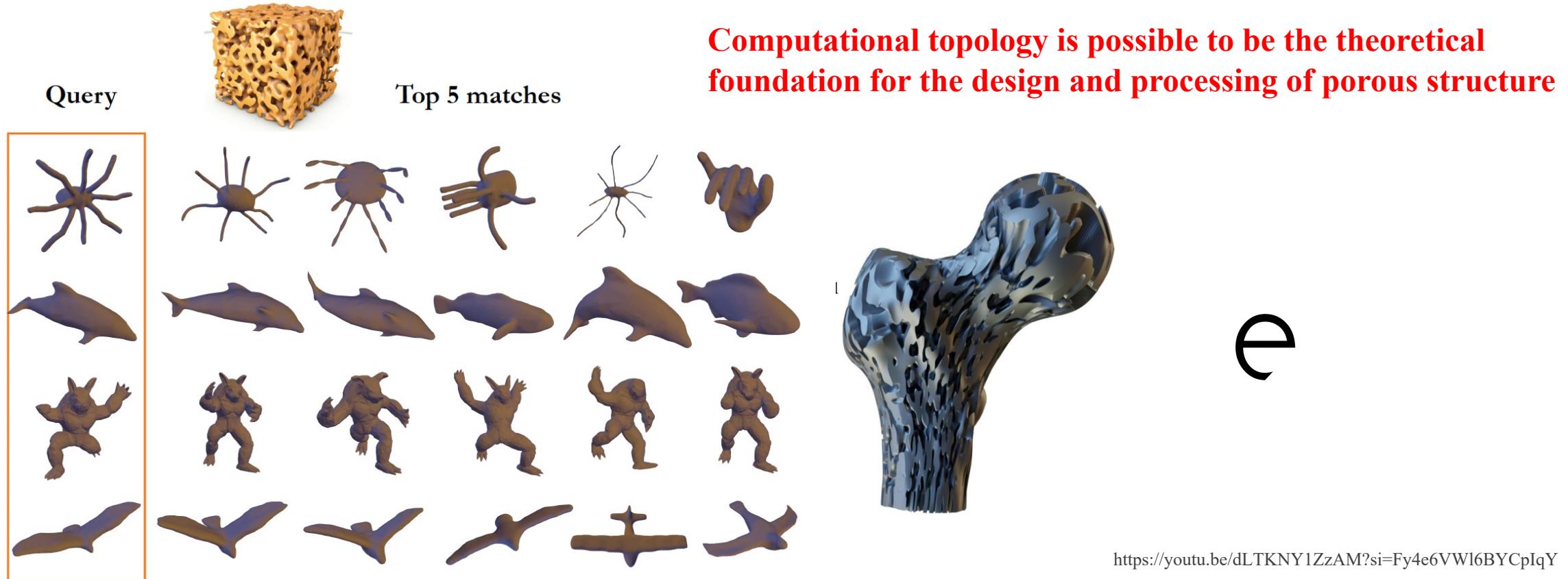
Yaqi He, Jiacong Yan, Hongwei Lin. Robust reconstruction of closed parametric curves by topological understanding with persistent homology. Computer-Aided Design, 165: 103611, 2023



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# Introduction

- **Porous Retrieval:** Given a query (3D model), retrieving models **most similar** to the query in a model base
- **Porous Design:** Keep **connectivity** of the designed porous
- **Porous Printing:** Guarantee **topology consistency** between design model and printing model

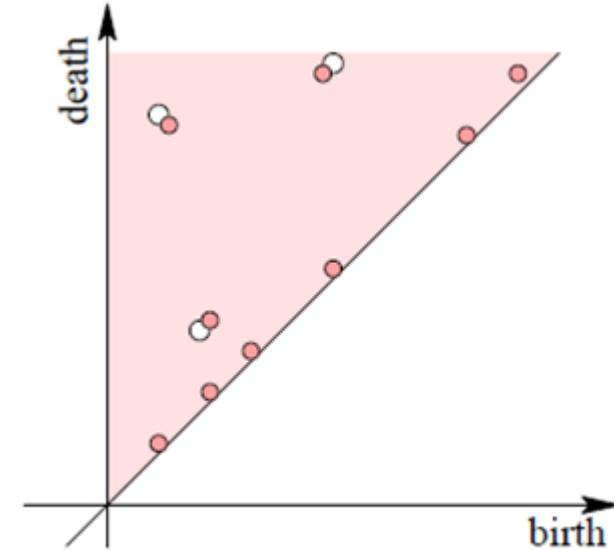


<https://youtu.be/dLTKNY1ZzAM?si=Fy4e6VWl6BYCpIqY>

- **Analysis and applications of PDs**
  - To extract more statistical information from PDs
  - To learn from PDs via machine learning approaches
  - To design feature descriptors using PDs
- **Space of PD with metric  $W_p$** 
  - Irregular points on a PD
  - Inefficient to compute the  $W_p$  distance

PD should be transformed into  
a proper representation.

Vectorizing Representation of a PD:  
to map a PD into a Hilbert space.



$$W_p(PD^{(1)}, PD^{(2)}) = \inf_b \left( \sum_{u \in PD^{(1)}} \|u - b(u)\|_\infty^p \right)^{1/p}$$

# Representation: Persistence Images

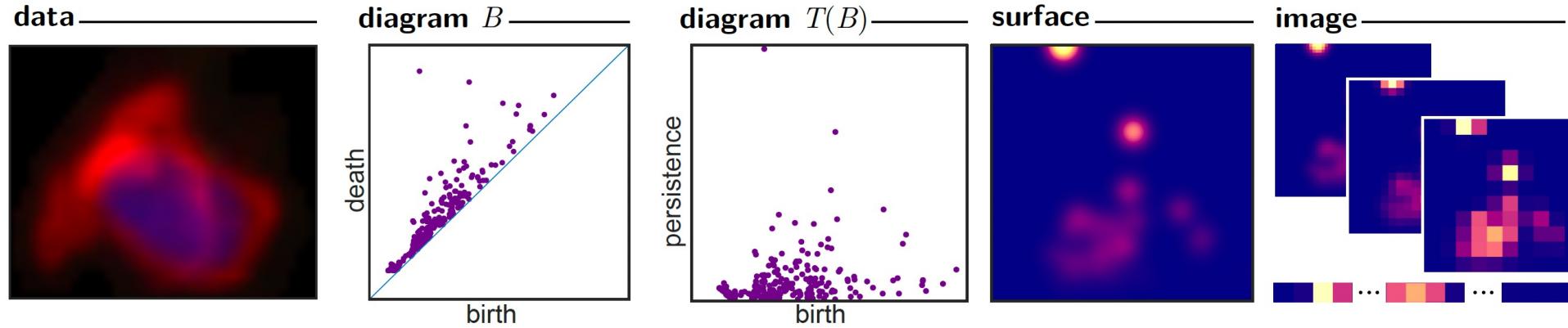
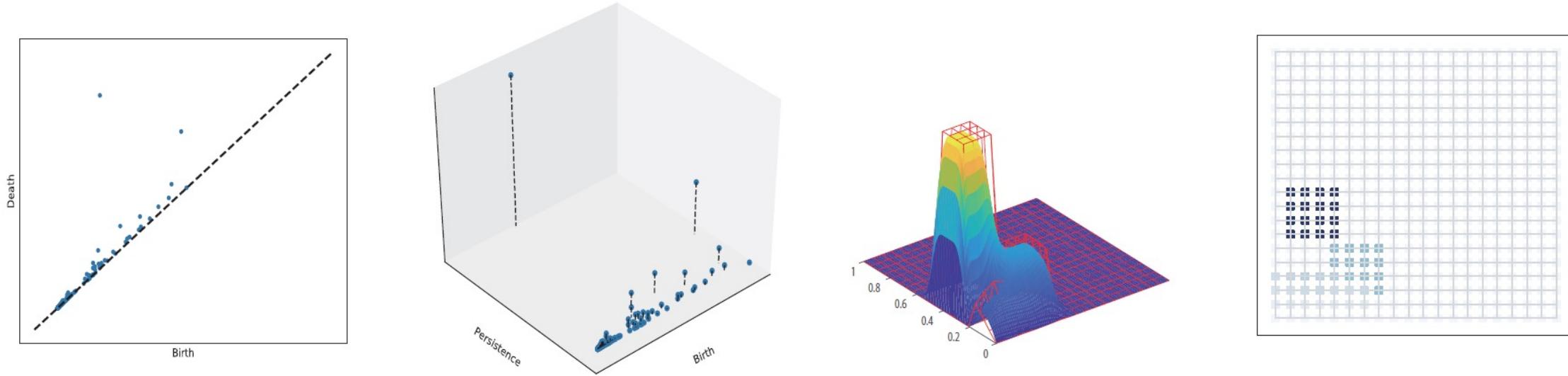


Figure 1: Algorithm pipeline to transform data into a persistence image.

- The **persistence image** (PI) provides a vector representation of the information in a persistence diagram
- The **Gaussian function** is chosen at each point in the transformed persistence diagram  $T(B)$
- It makes the results of persistent homology **available in machine learning** algorithms such as linear SVM

Adams H, Emerson T, Kirby M, et al. Persistence images: A stable vector representation of persistent homology. Journal of Machine Learning Research, 2017, 18(8): 1-35.

# Vectorization: Persistence B-spline Grid (PBSG)



- Transform a PD into the birth-persistence coordinates and assign a value to each point in a PD according to its importance
- Fit the data points in 3D by a **cubic uniform B-spline surface**
- The control grid is obtained to **generate a vector** by concatenating rows of z-coordinates of the control grid
- **Persistence B-spline Grid** is defined by the matrix formed by z-coordinates of the control points

**PBSG performs better than PI in most cases**

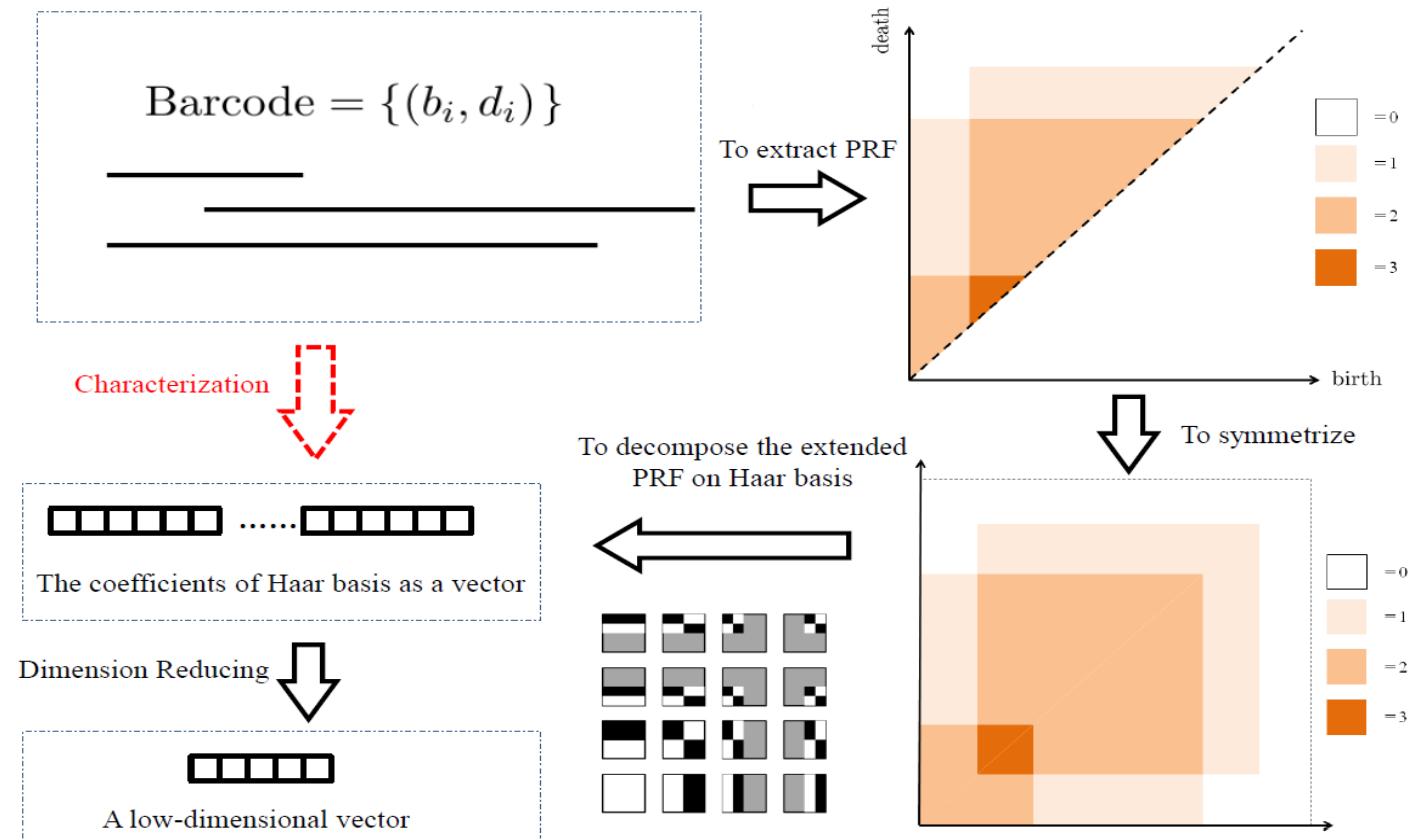
Zhetong Dong, Hongwei Lin, Chi Zhou, Ben Zhang, Gengchen Li. Persistence B-Spline Grids: Stable Vector Representation of Persistence Diagrams Based on Data Fitting. Machine Learning, 113(3): 1373-1420, 2024

# Characterization of Persistence Barcode by Haar Basis

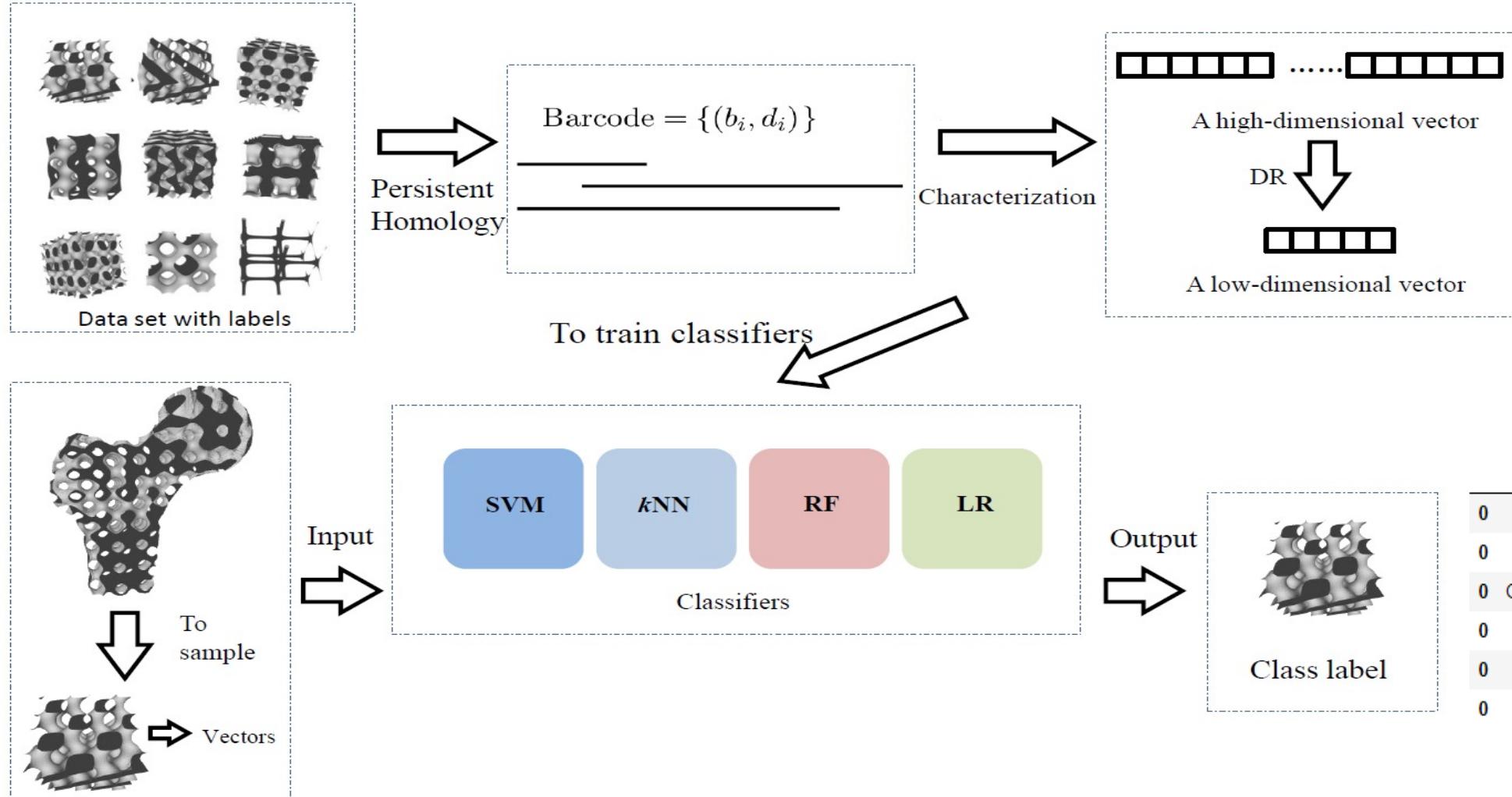
- Transform Barcode to **extended PRF**
- Wavelet decomposition on extended PRF
- Generate **shape descriptor** by dimension reduction to the wavelet decomposition coefficients

- Given Barcode  $\{(b_i, d_i) | 0 \leq b_i \leq d_i, i \in I, |I| < \infty\}$
- Extended persistent rank function** is defined as:

$$\tilde{r}_i(s, t) = \begin{cases} 1, & b_i \leq s, t \leq d_i \\ 0, & \text{otherwise} \end{cases}$$



# Characterization of Persistence Barcode by Haar Basis

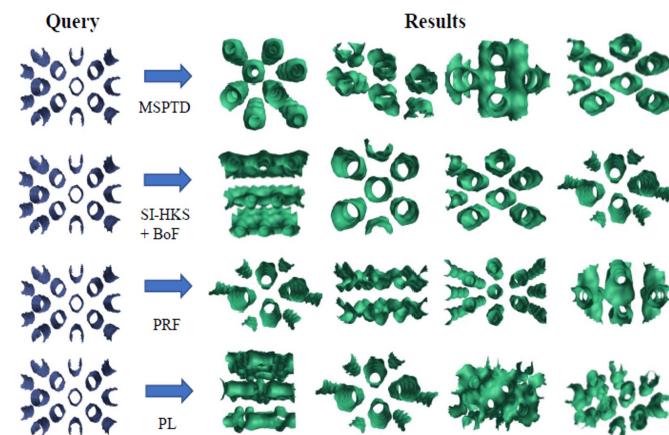
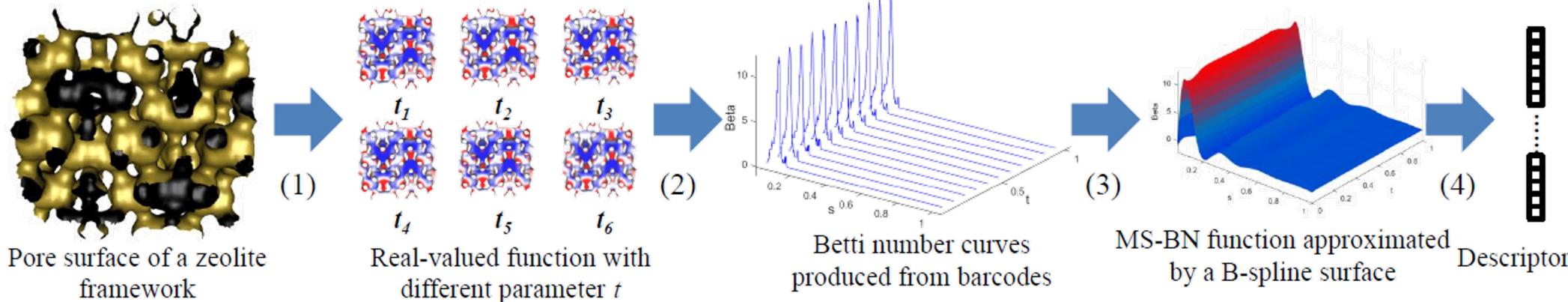


	Classifier	Accuracy
0	KNeighborsClassifier	0.990000
0	RandomForestClassifier	0.986111
0	GradientBoostingClassifier	0.985000
0	LogisticRegression	0.982222
0	LinearSVC	0.981667
0	SVC	0.990000

Zhetong Dong, Chuanfeng Hu, Chi Zhou, Hongwei Lin, Vectorization of persistence barcode with applications in pattern classification of porous structures , Computers & Graphics, Volume 90, 2020,Pages 182-192

# Multiscale Persistent Topological Descriptor

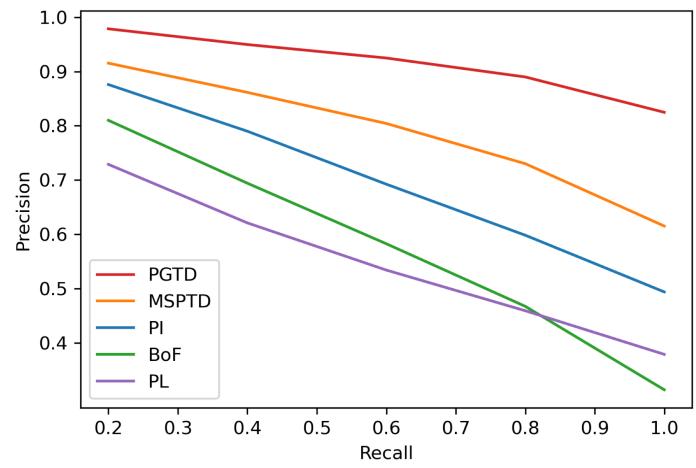
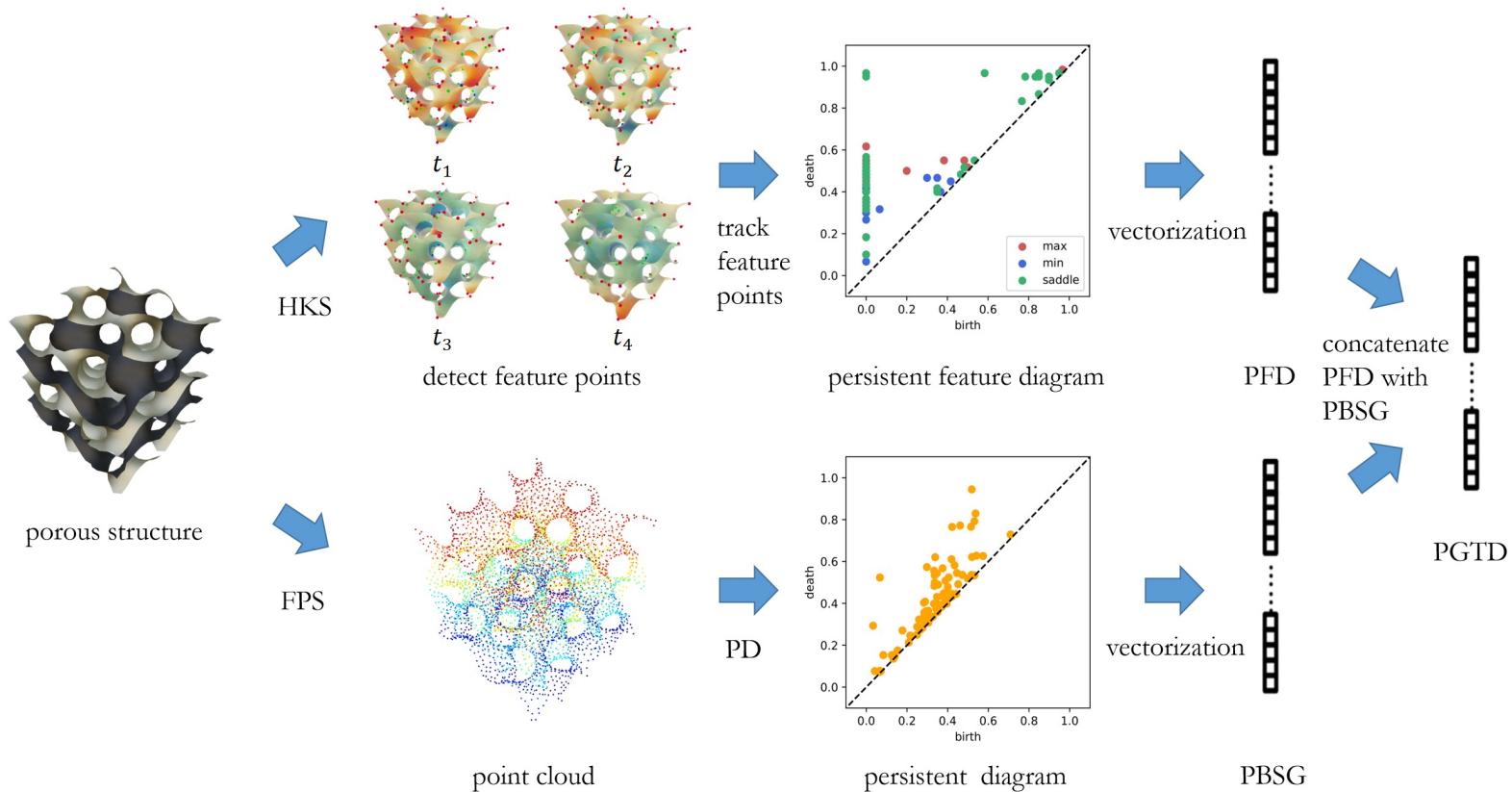
- Calculating a series of **Betti number curves** of the heat kernel on a porous surface at some times
- Generating the **B-spline surface** by lofting the Betti number curve sequence
- The **control grid of the B-surface surface** is taken as the **shape descriptor**



Methods/Accuracy (%)	NN	FT	ST	DCG
SI-HKS with histogram	90.1	73.3	91.6	92.7
SI-HKS with BoF	81.3	76.1	91.0	91.8
PL descriptor	68.1	59.3	75.0	83.4
PI descriptor	81.5	71.8	86.7	90.1
PRF descriptor	81.8	74.1	87.5	90.6
MSPTD (ours)	<b>94.9</b>	<b>78.4</b>	<b>94.5</b>	<b>95.5</b>

# Persistent Geometry-Topology Descriptor

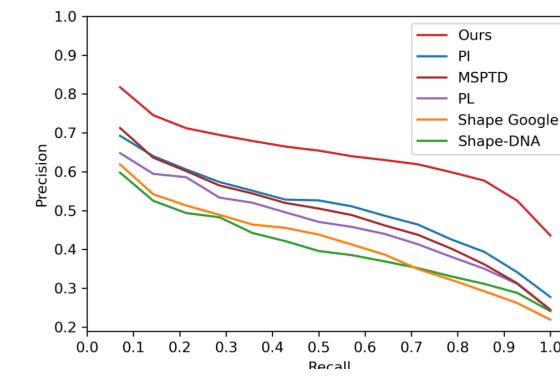
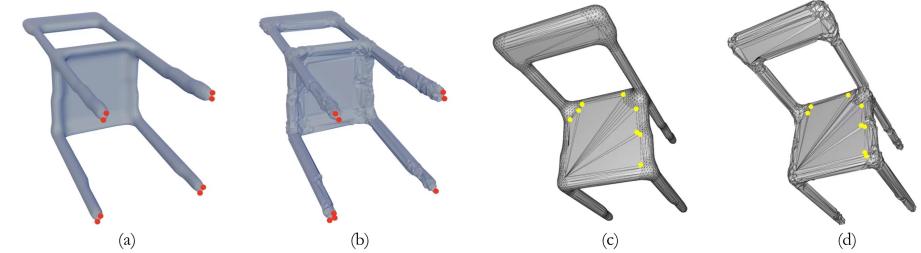
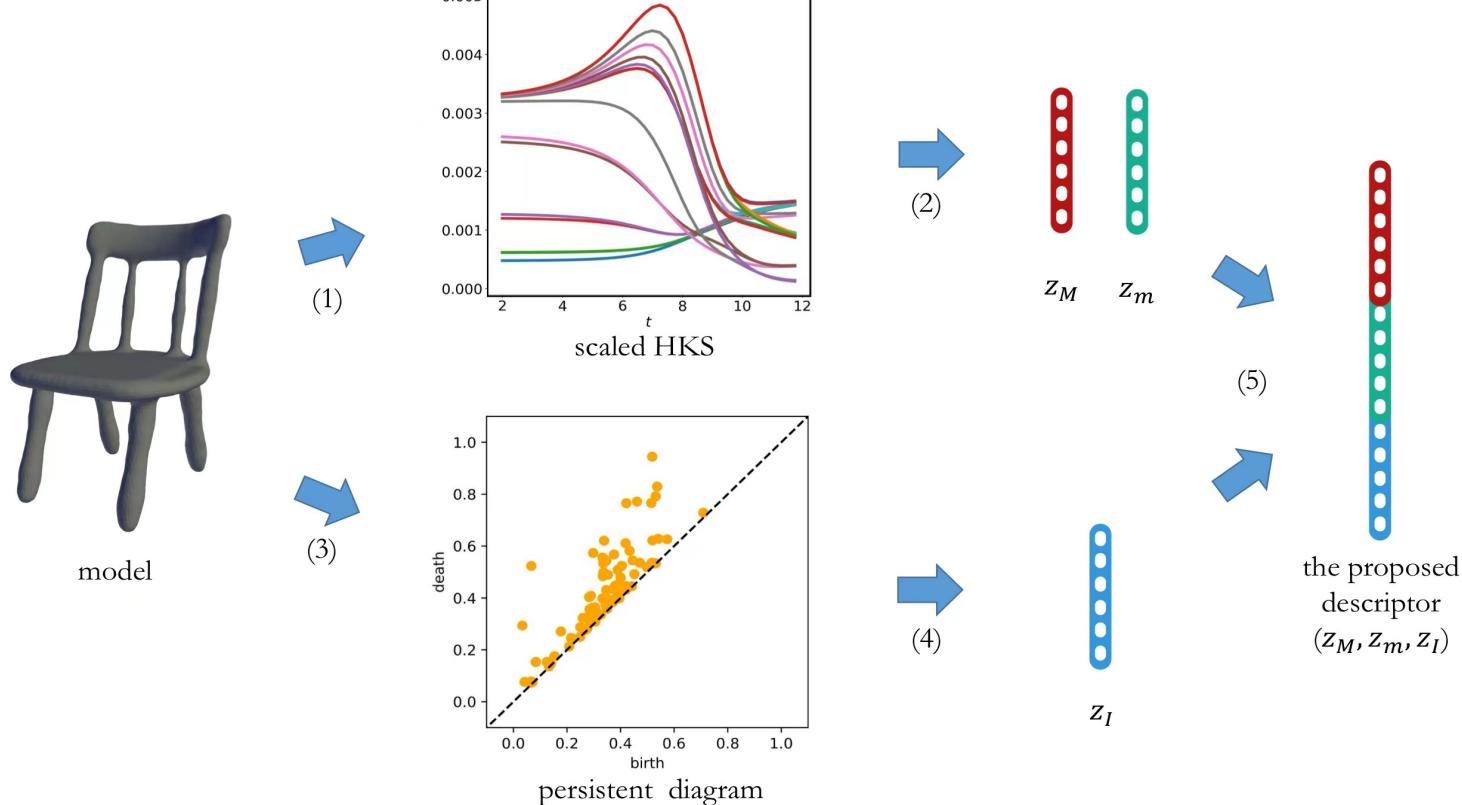
- Geometric information is represented by **persistent feature diagram (PFD)**, formed by tracing the emergence and disappear of the heat kernel feature points
- Topological information is represented by **persistent diagram (PD)**
- Persistent Geometry-Topology Descriptor** is generated by concatenating the vectors of PFD and PD



Methods/Accuracy (%)	NN	FT
MSPTD	86.8	71.0
HKS with BoF	75.1	50.8
PL descriptor	62.8	48.9
PI descriptor	81.6	61.8
<b>PGTD (ours)</b>	<b>96.3</b>	<b>86.5</b>

# Persistent Heat Kernel Signature Descriptor

- Calculate the **heat kernel signature curve** at each mesh vertex, as well as the **maximums and minimums** on the curves
- Shape descriptor** is formed by first sorting the maximums and minimums, and then concatenating them with the vector of PD

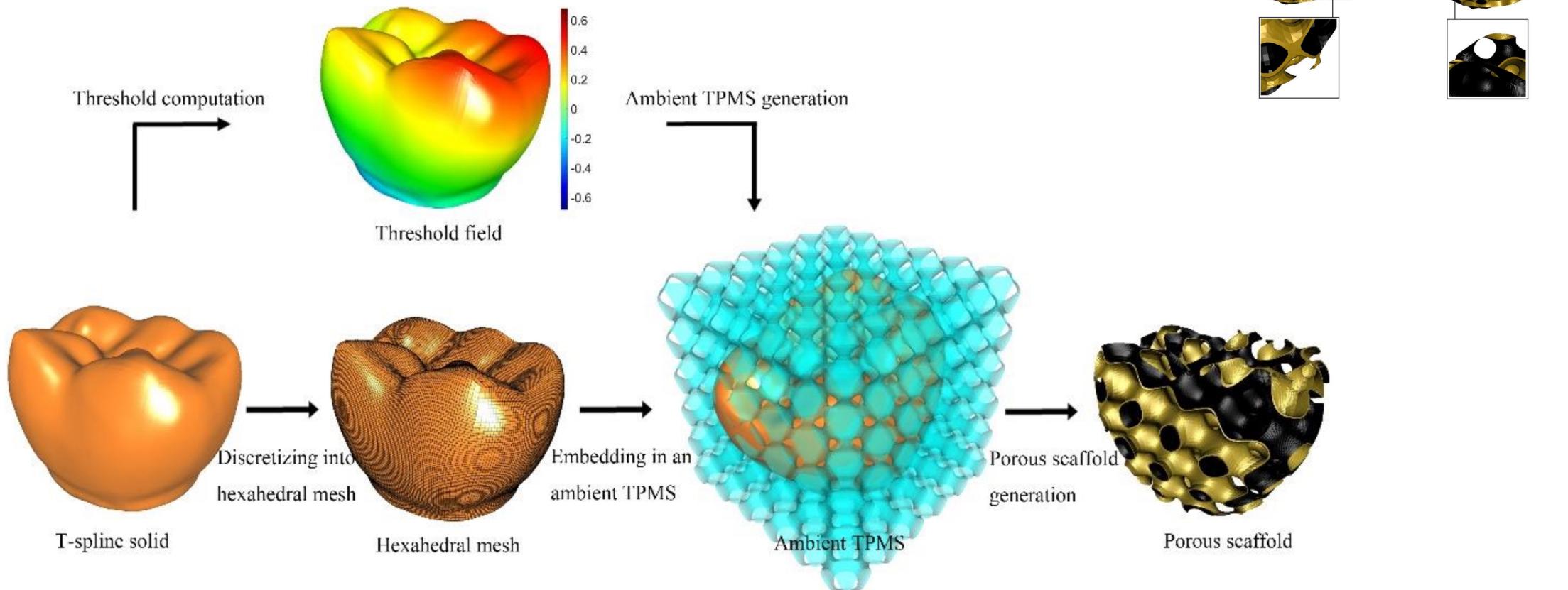


Methods/Accuracy (%)	NN	FT	ST
Shape-DNA (Reuter et al., 2006)	46.49	37.66	53.01
Shape Google (Bronstein et al., 2011)	52.63	37.34	52.57
PI (Adams et al., 2017)	61.40	44.67	62.47
PL (Bubenik, 2015)	57.89	40.85	58.02
MSPTD (Dong et al., 2021)	60.53	43.92	62.66
Our descriptor	<b>78.07</b>	<b>59.09</b>	<b>73.56</b>

Zitong He, Peisheng Zhuo, Hongwei Lin, Junfei Dai. 3D shape descriptor design based on HKS and persistent homology with stability analysis. Computer Aided Geometric Design, 111 (2024) 102326

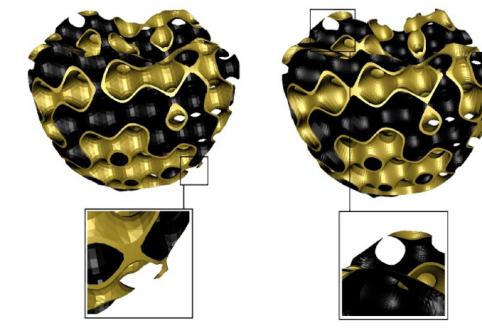
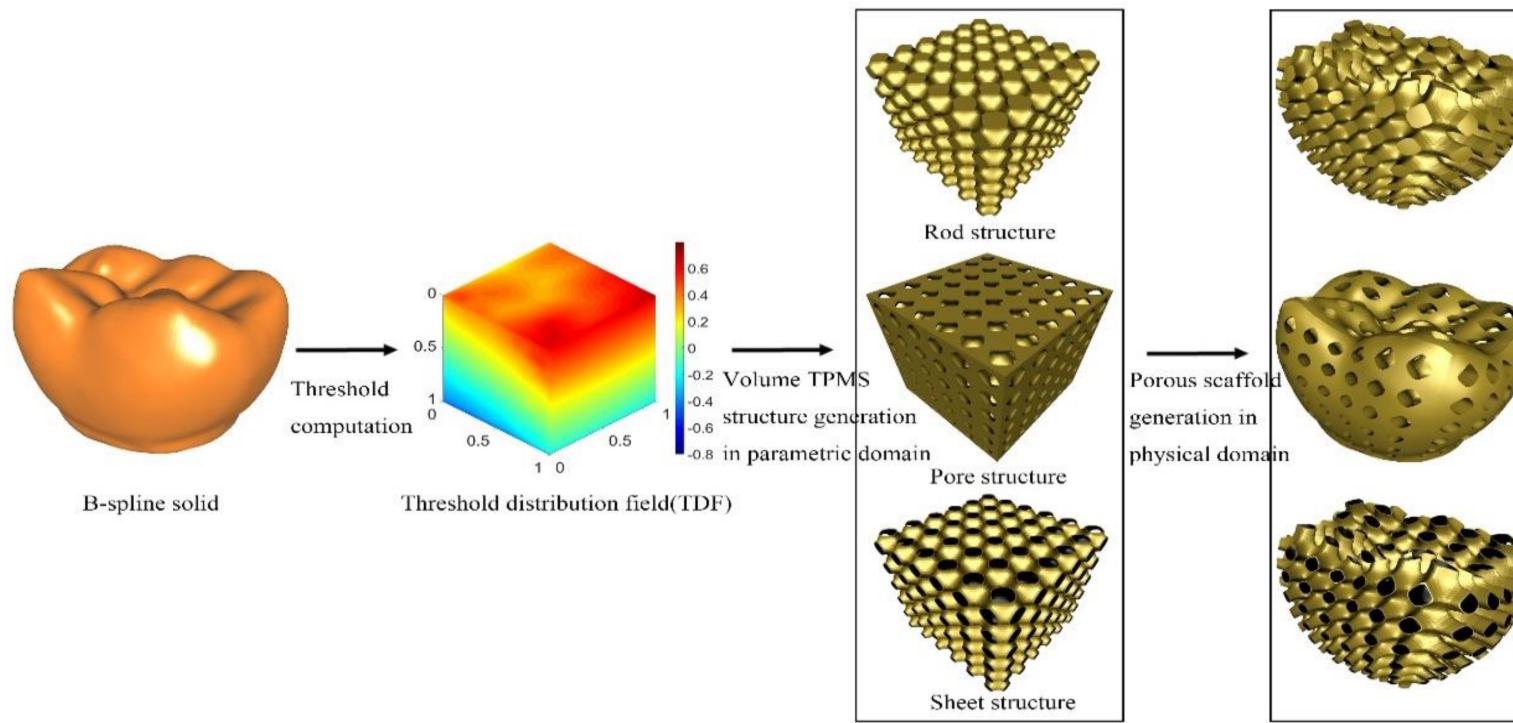
# Heterogeneous porous : Conventional Methods

- Conventional porous generation methods



# Heterogeneous porous : Our Methods

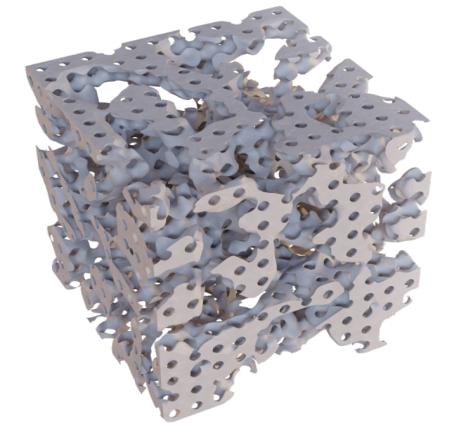
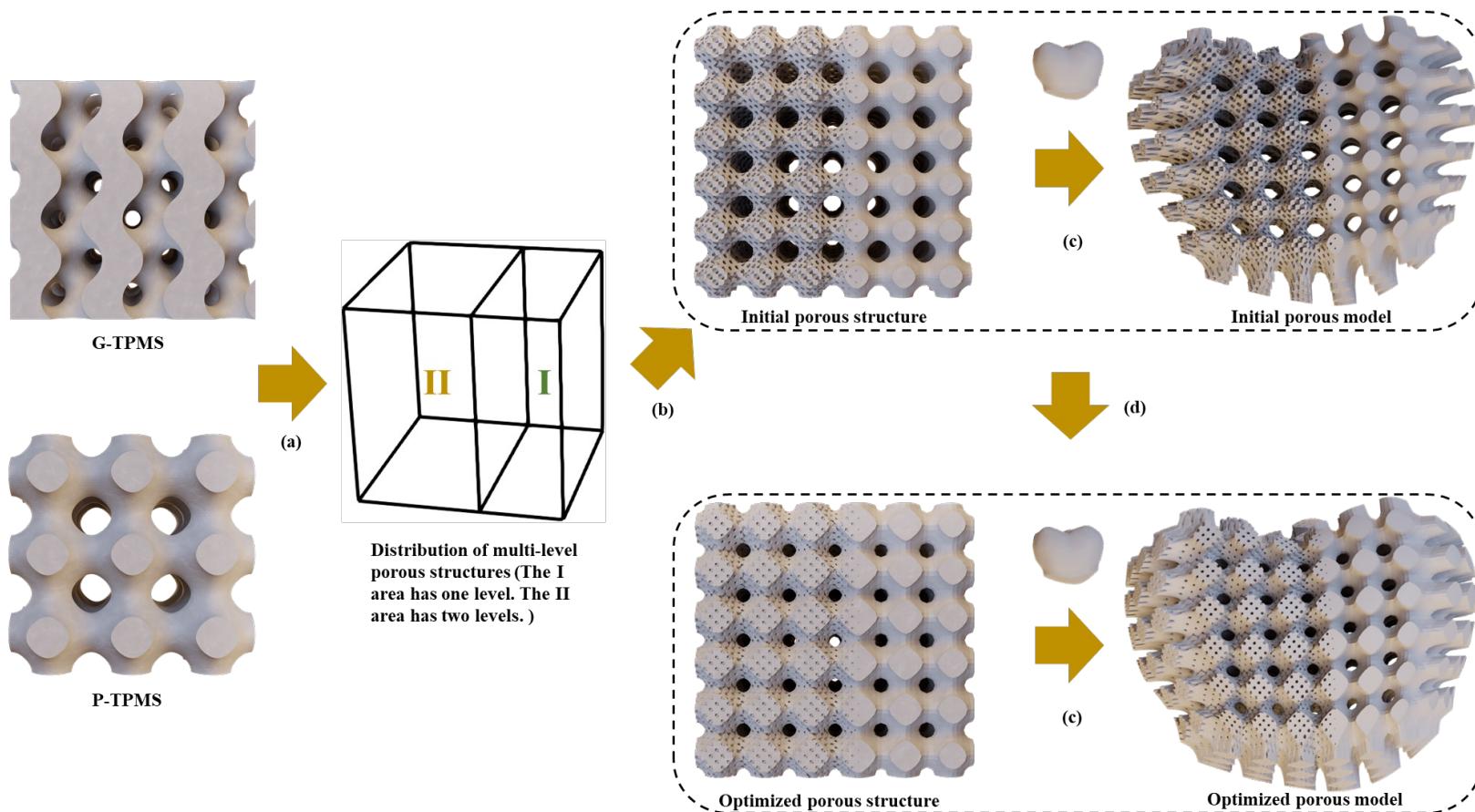
- Our method: The **composition** of the free form shape represented by **tri-variate B-spline** and the **implicit B-spline** in the parametric domain
- Save at least **99%** storage
- Surface integrity



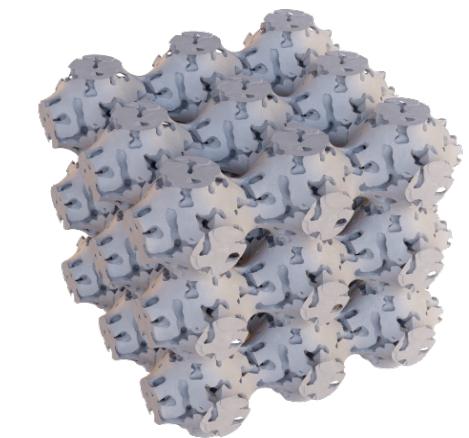
Chuanfeng Hu, Hongwei Lin. Heterogeneous porous scaffold generation using trivariate B-spline solids and triply periodic minimal surfaces. Graphical Model. 115: 101105, 2021

# Multilevel porous structure: Algorithm flow

- Represent multilevel porous structure by **hierarchical B-spline**
- Implicit B-spline** is employed to represent porous at each level parametric domain

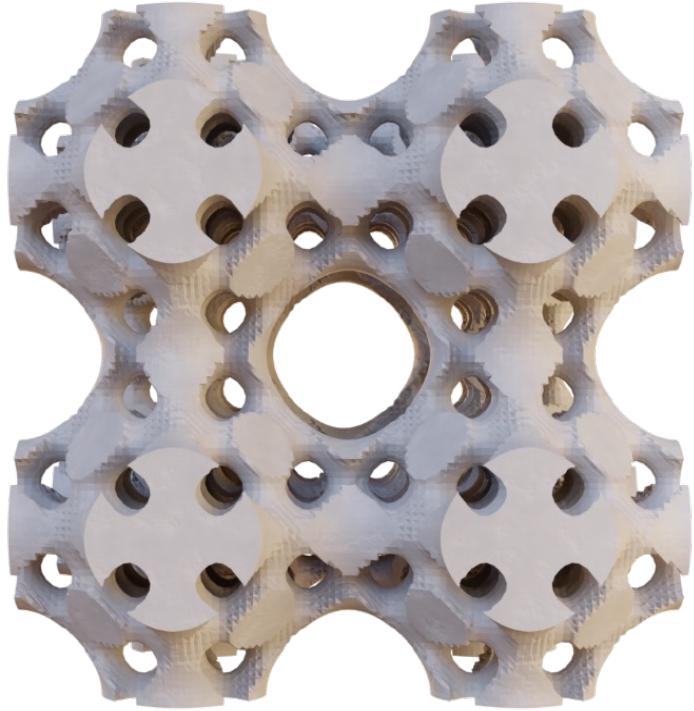


Bionic porous in level-0  
and P-TPMS in level-1.

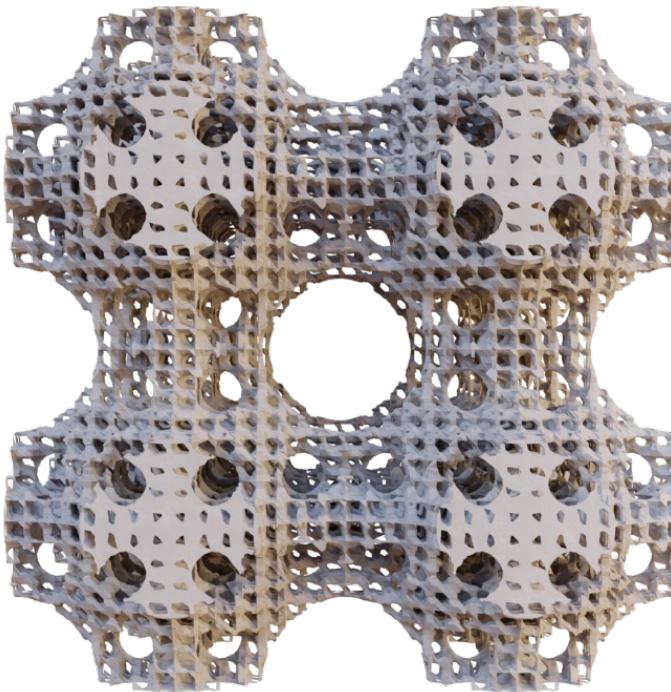


P-TPMS in level-0 and Bionic  
porous in level-1.

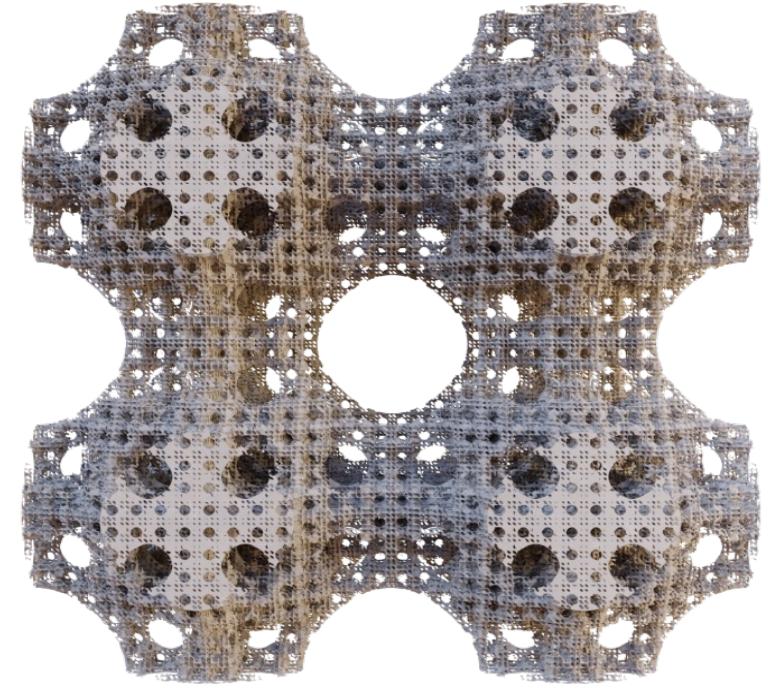
# Multilevel porous structure: Examples



Two level



Three level

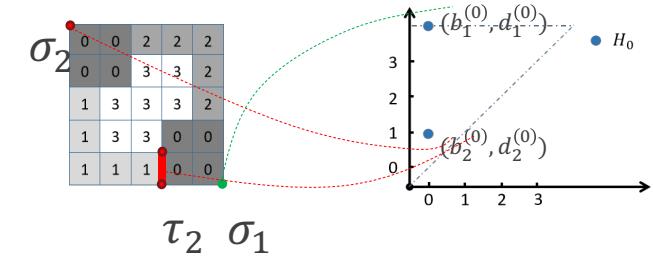
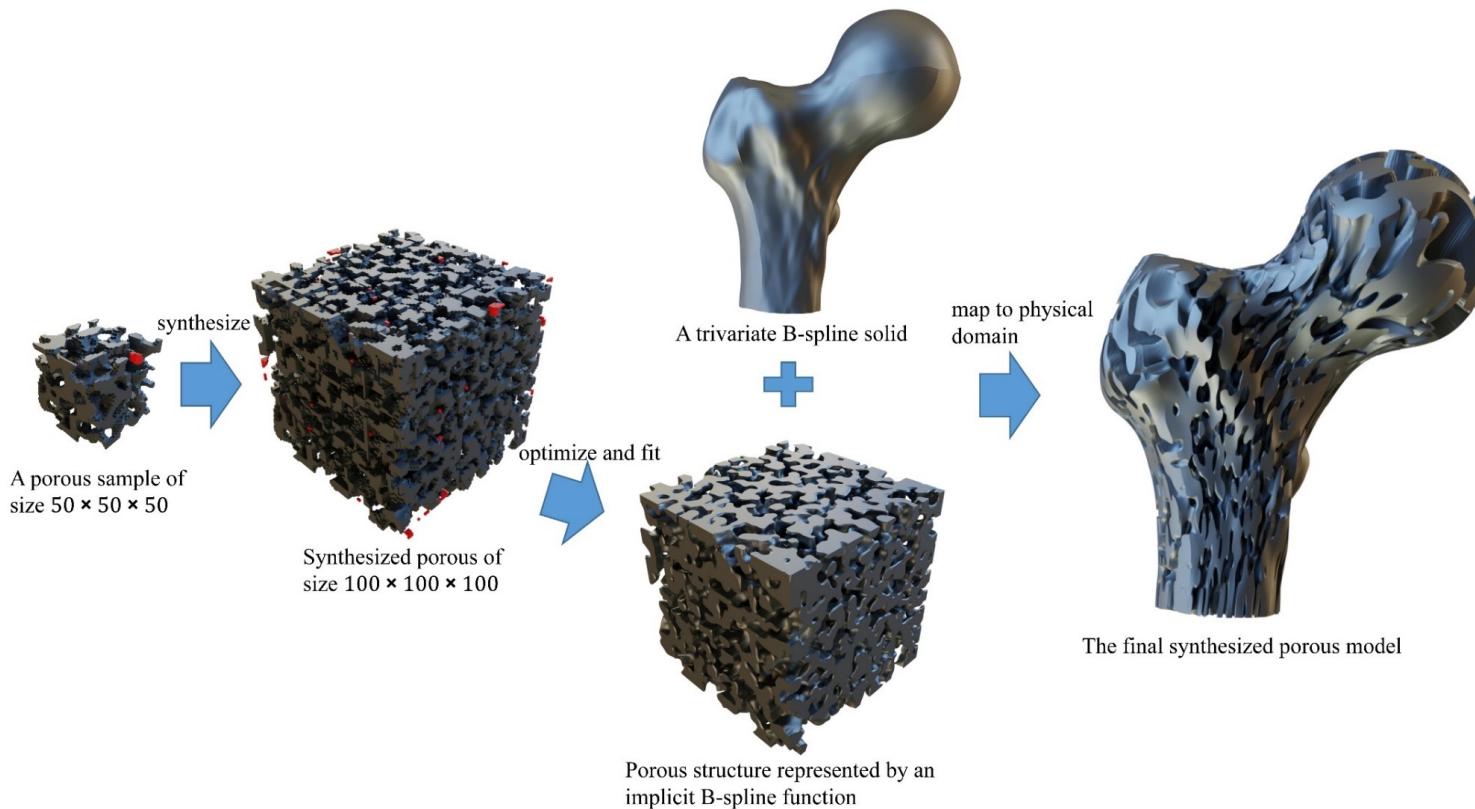


Four level

Depeng Gao, Hongwei Lin, Zibin Li. Free-form multi-level porous model design based on truncated hierarchical B-spline functions. Computer-Aided Design, 2023, 162: 103549

# Connectivity-guaranteed porous synthesis

- Synthesize the porous in a large region from a small region porous sample using texture synthesis method
- Guarantee the connectivity of the synthesized porous by optimizing the PD

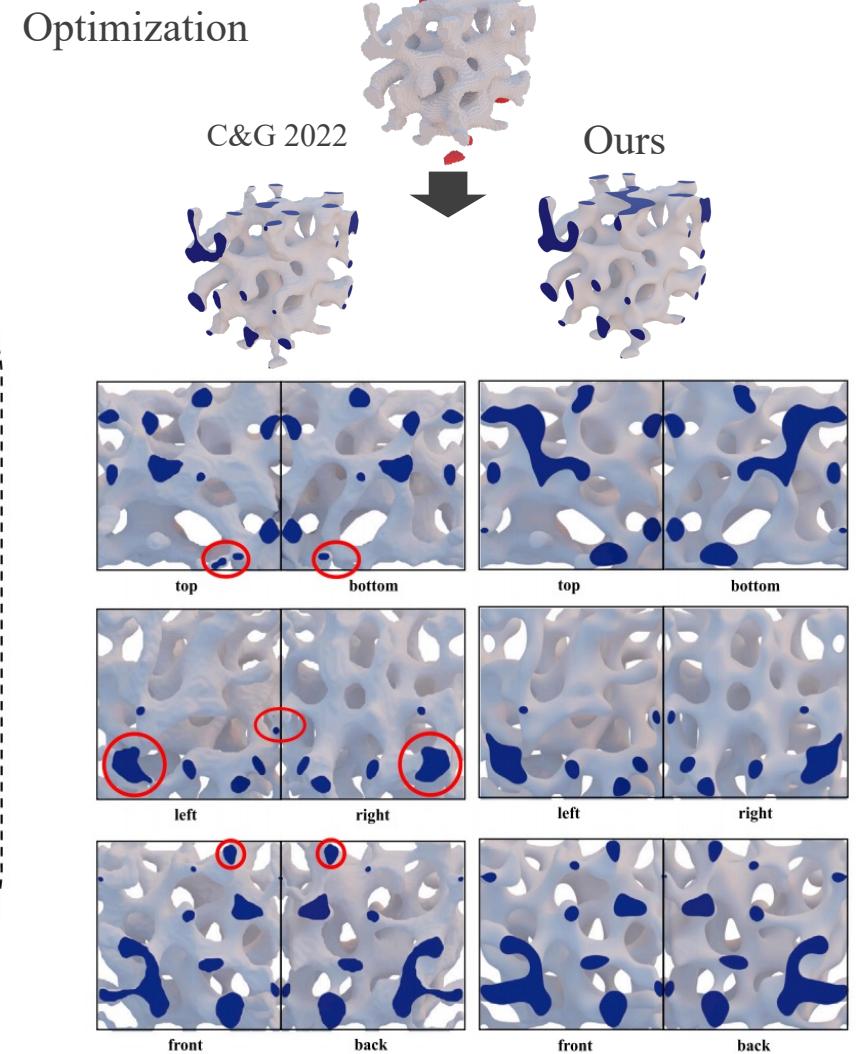
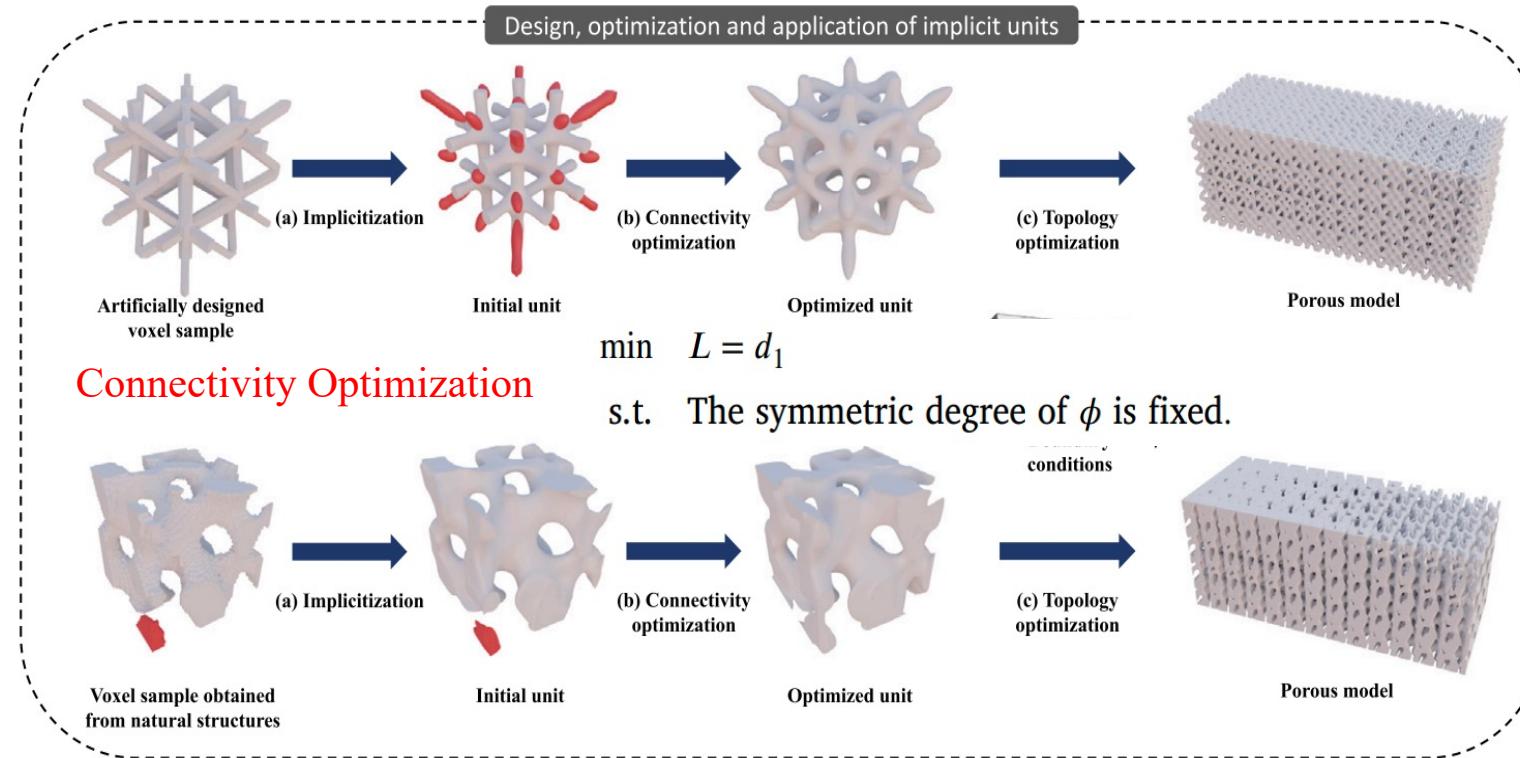


- DTM based filtration
- PD optimization

Depeng Gao, Jinhao Chen, Zhetong Dong, Hongwei Lin. Connectivity-guaranteed porous synthesis in free form model by persistent homology. Computers & Graphics, 106: 33-44 (2022)

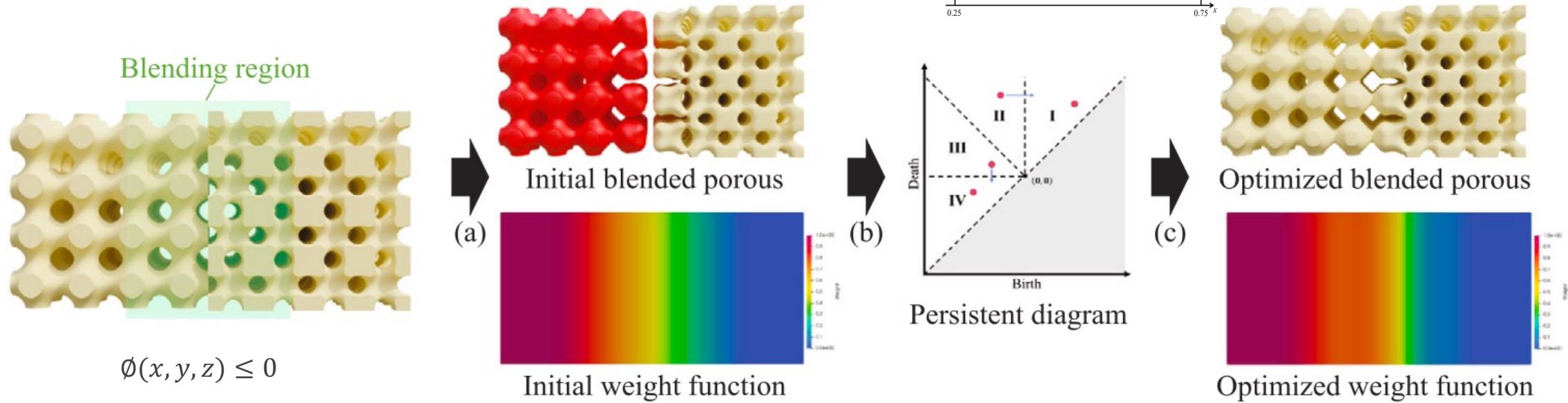
# Periodic Implicit Representation: Flowchart

- Represent the porous unit by **periodic implicit B-spline**
- The porous units can be stitched watertight and smoothly
- Enrich the representation forms and design methods of porous structures



# Topology-aware blending method

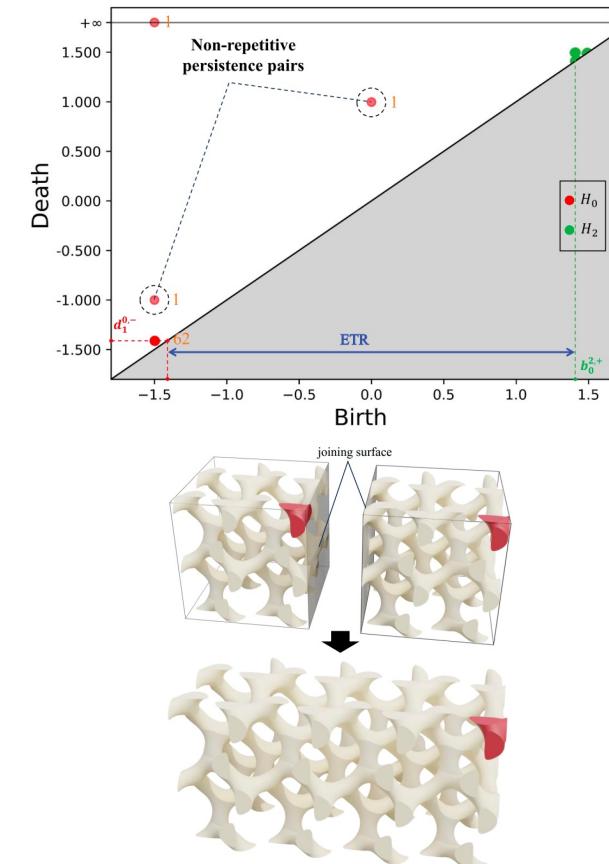
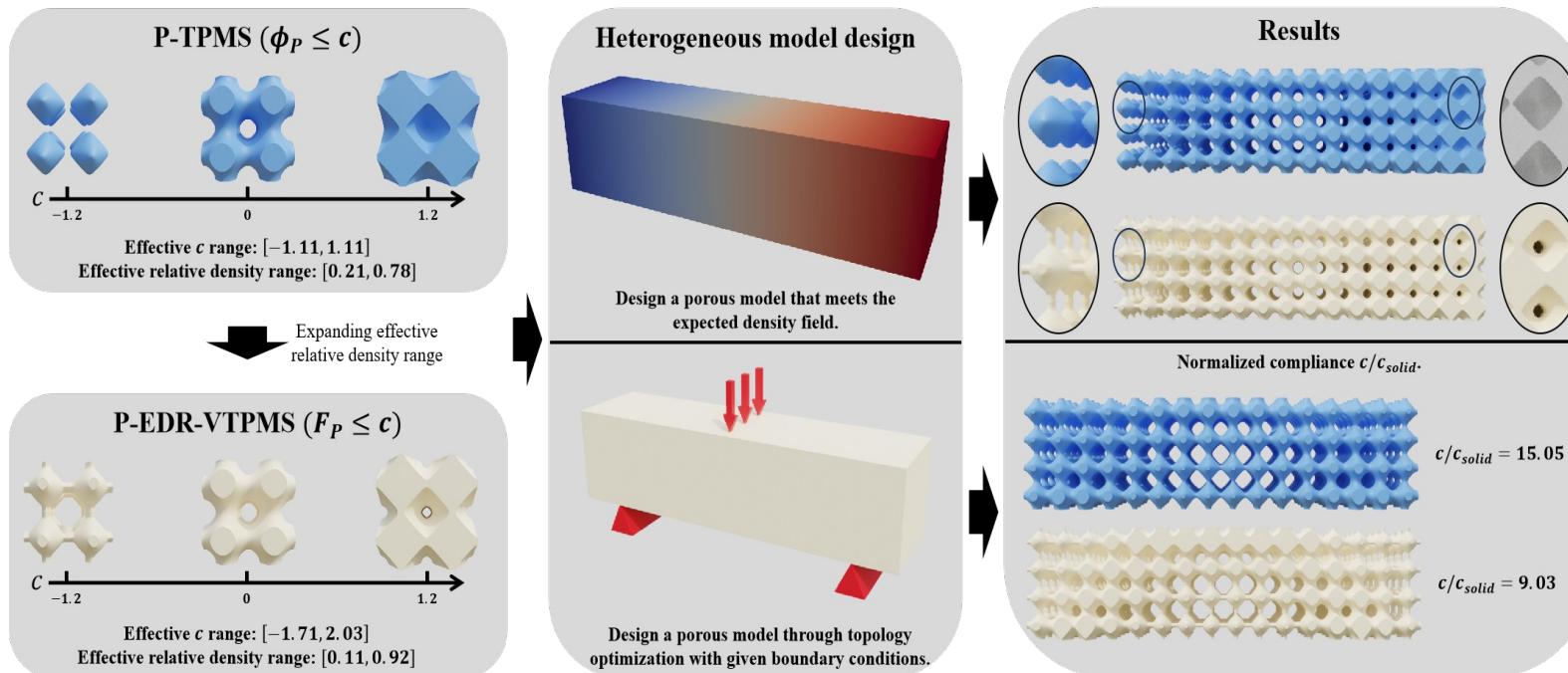
- Blend two implicitly represented porous structures smoothly
- Eliminate the topological errors in the blending region by PD optimization
- Keep the porous outside the blending region unchanged



Depeng Gao, Yang Gao, Yuanzhi Zhang, Hongwei Lin. Topology-aware blending method for implicit heterogeneous porous model design. Computer Aided design, 177:103782 (2024)

# Persistent Homology-Driven Optimization of Effective Relative Density Range

- Traditional porous representation is TPMS  $\Phi(x) < C$
- $\Phi(x)$  is a trigonometric function, with a little adjustable parameters, and limited valid range of  $C$
- Replace  $\Phi(x)$  with B-spline, and enlarge the valid range of  $C$  by **PD optimization**

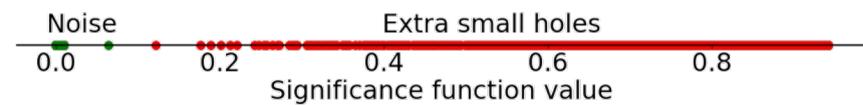
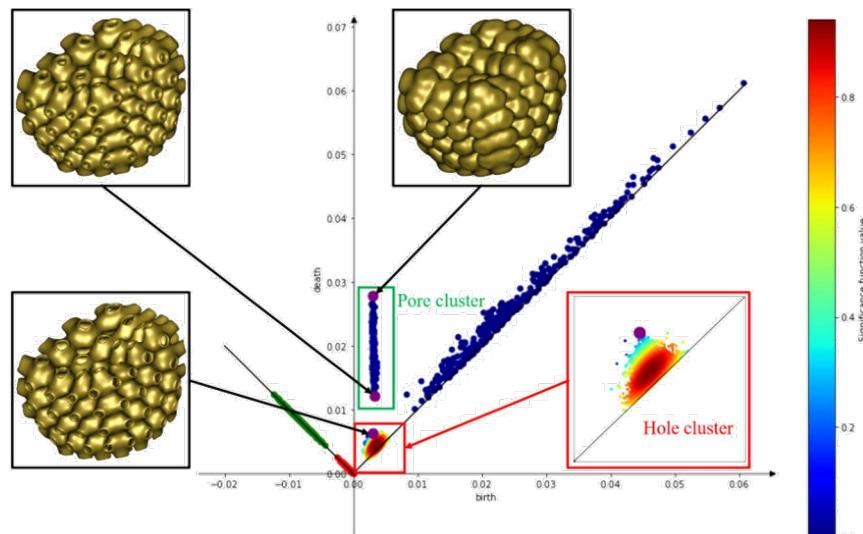


EDR-VTPMS: Effective relative density range-Variant TPMS

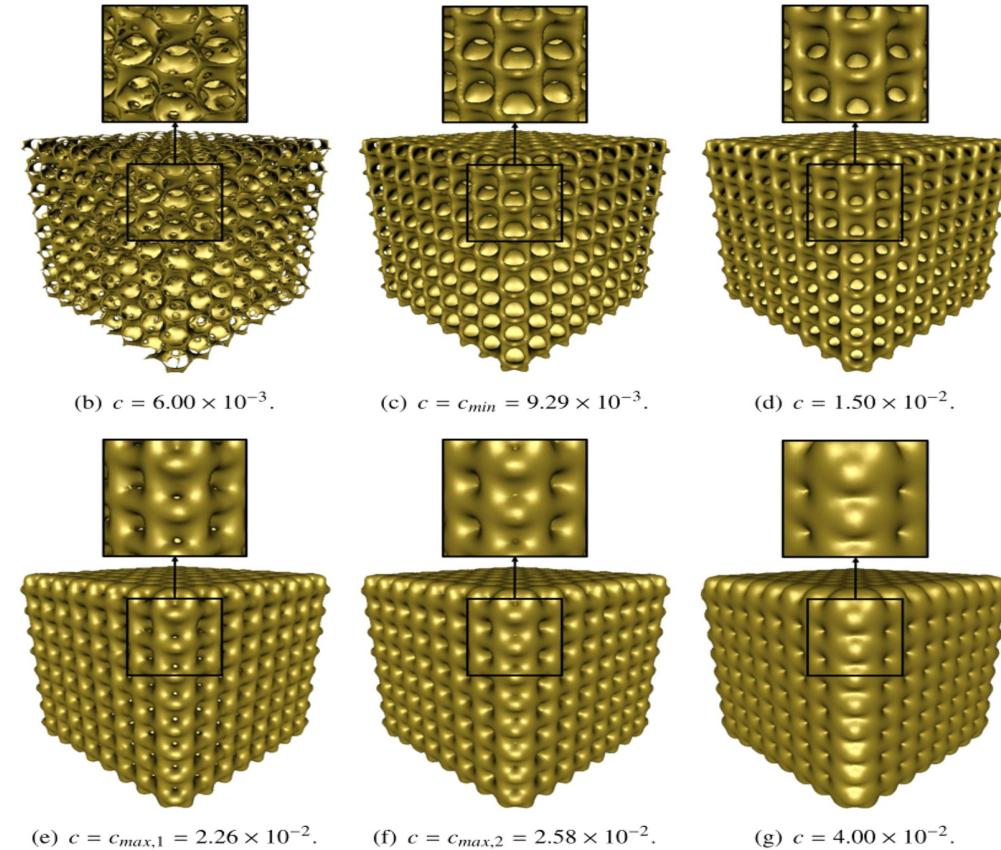
Depeng Gao, Yuanzhi Zhang, Hongwei Lin. Persistent Homology-Driven Optimization of Effective Relative Density Range for Triply Periodic Minimal Surface. Computer-Aided Design, in revision.

# Reasonable thickness determination for sheet structure

- TPMS surface represented porous should have thickness for 3D
- With too small thickness, there will be **extra small holes**; with too large thickness, porous will be **closed**
- By **clustering the points in a PD**, determine the reasonable thickness range

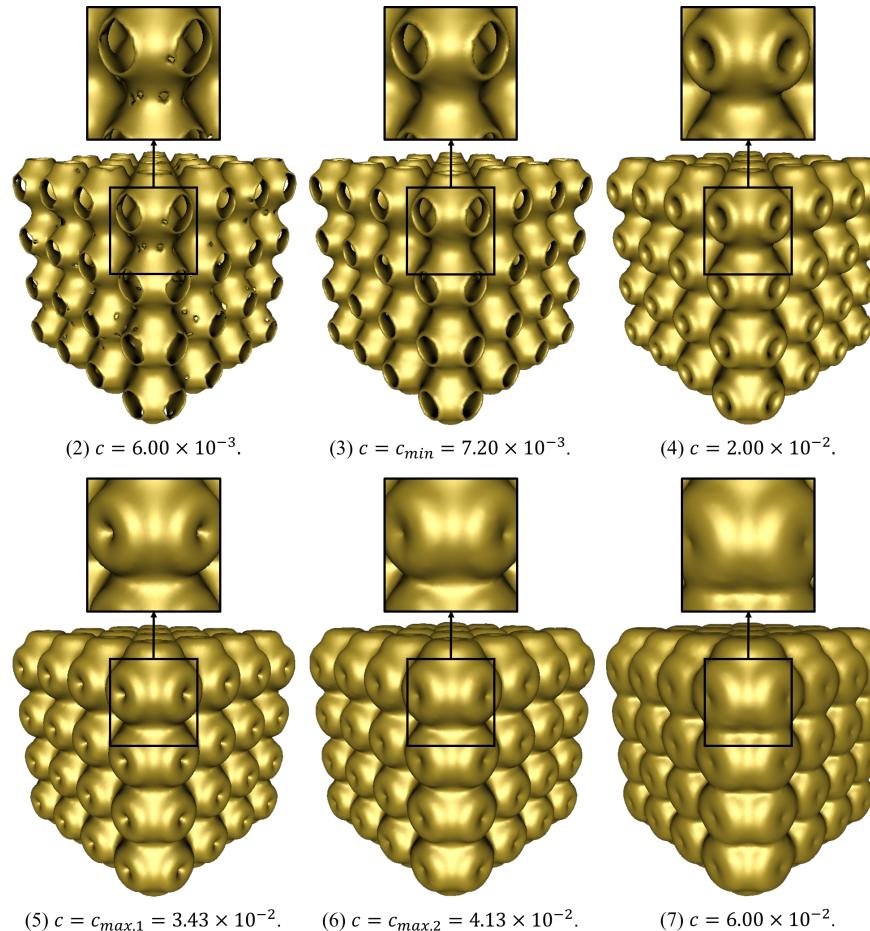
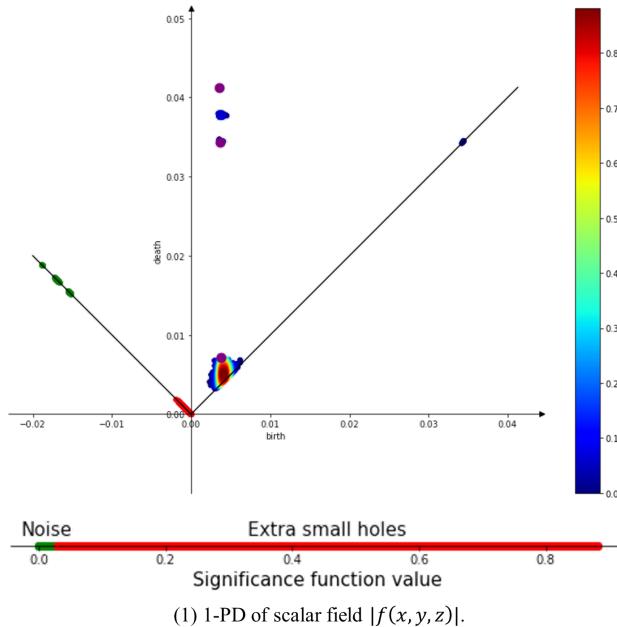


(b) Significance function values of the points in the short persistence cluster.



# Reasonable thickness determination for porous sheet structure

## Experimental Results:



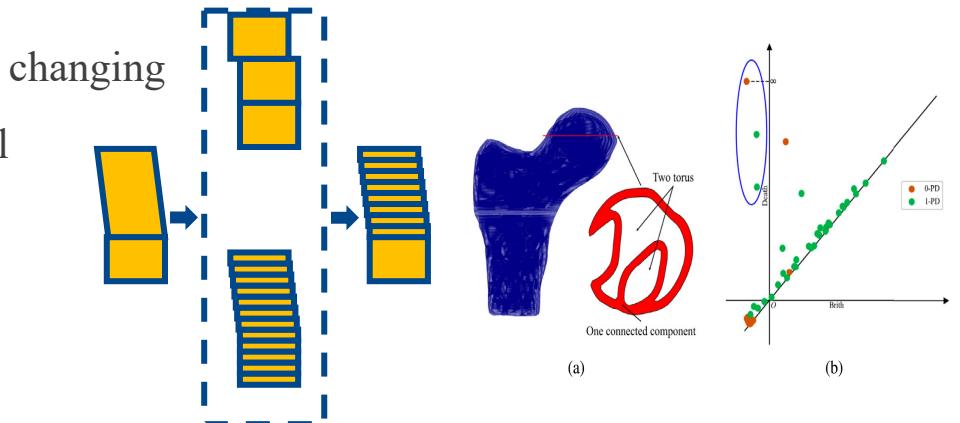
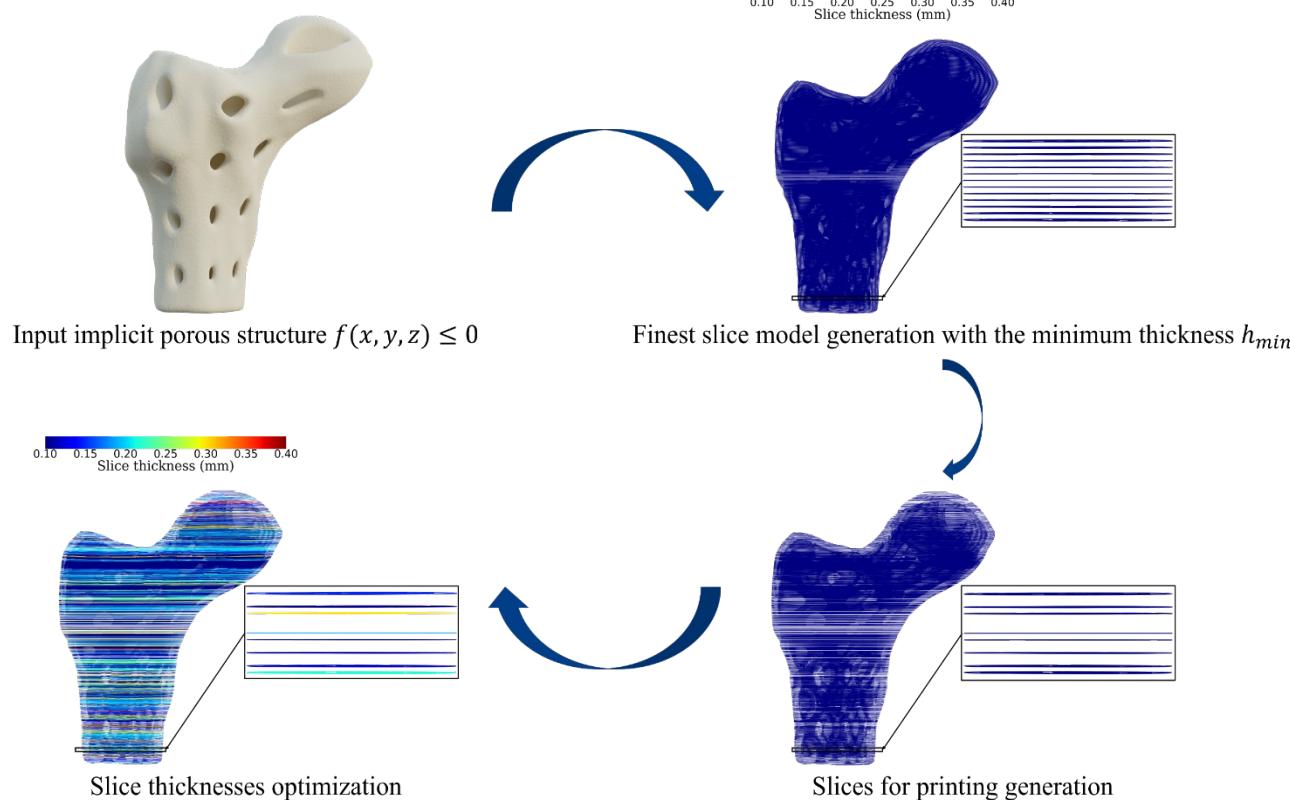
The variation of the number of pores and holes in these sheet structures

Parameter $c$	Topological measurement $\bar{\beta}$
$c < c_{min}$	5.661
$c = c_{min}$	<b>1.000</b>
$c_{min} < c < c_{max,1}$	<b>1.000</b>
$c = c_{max,1}$	0.998
$c = c_{max,2}$	0.000
$c > c_{max,2}$	0.000

Jiacong Yan, Hongwei Lin. Reasonable thickness determination for implicit porous sheet structure using persistent homology. Computers & Graphics, 115, 236-245, 2023

# Adaptive slicing with topology guarantee

- The 2D slice of a porous has **complicated topological structure**
- By **comparing the PDs of adjacent slices**, find the critical slice with topology changing
- Guarantee the topology consistency** between design model and printed model



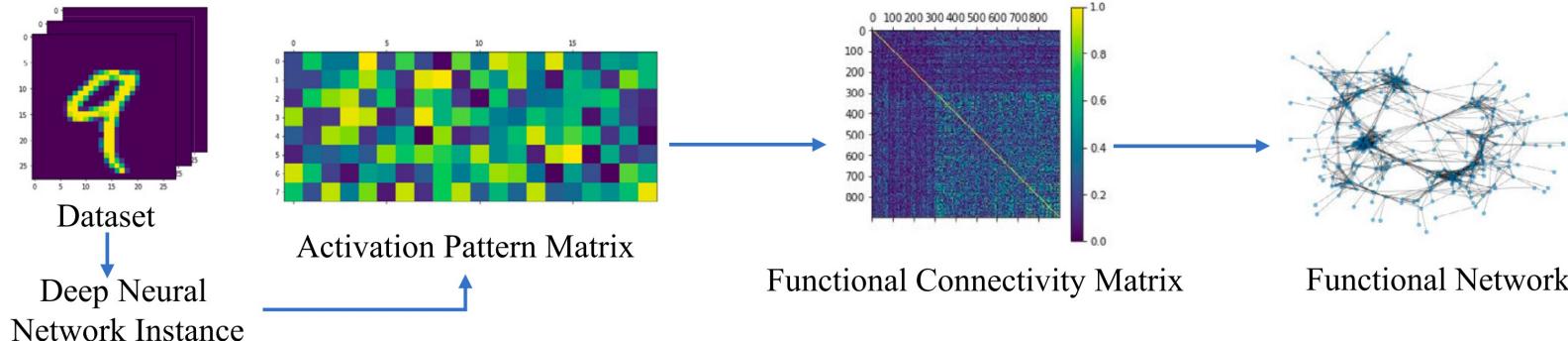
Model	Method	Slice number	Topology detection error	Cusp height sum (mm)	Volume deviation
Balljoint	Our method	502	<b>0</b>	44.34	0.31%
	Cusp height based method	502	192	47.52	0.39%
	Boolean operations based method	502	197	46.34	0.40%
Moai	Our method	580	<b>0</b>	35.56	0.40%
	Cusp height based method	580	192	37.45	0.41%
	Boolean operations based method	580	148	37.02	0.41%
Tooth	Our method	652	<b>0</b>	54.63	0.58%
	Cusp height based method	652	274	56.84	0.62%
	Boolean operations based method	652	233	55.93	0.62%
Venus	Our method	569	<b>0</b>	51.31	0.55%
	Cusp height based method	569	227	54.23	0.60%
	Boolean operations based method	569	264	53.20	0.61%



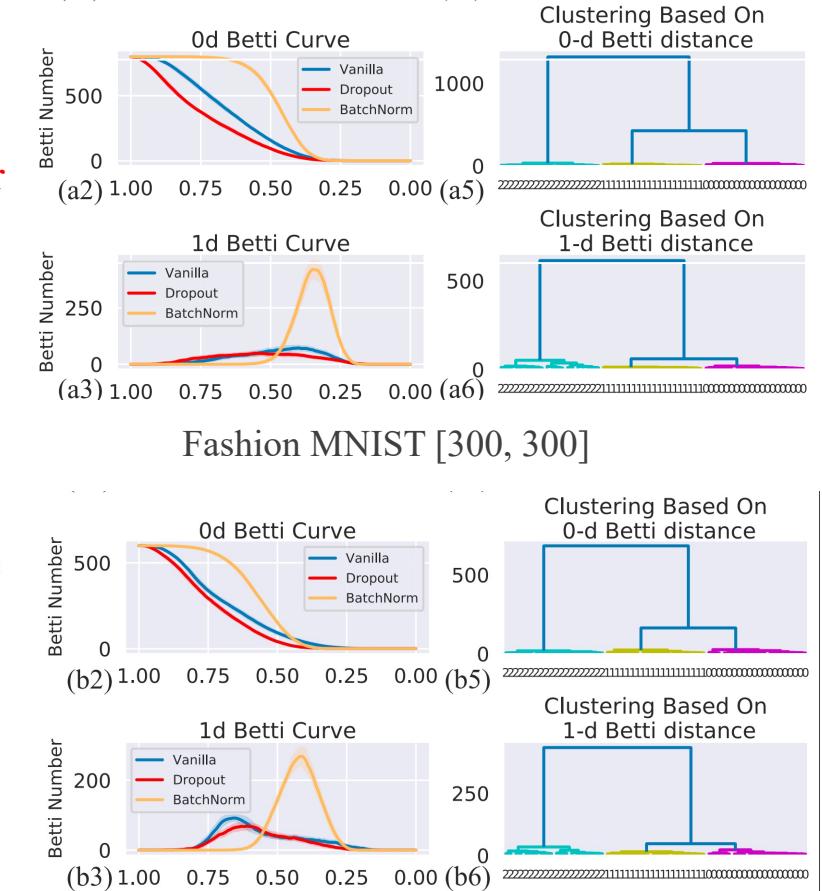
1. Introduction to computational topology
2. Applications in geometric design
3. Porous retrieval, design, and printing
4. Interpretability of DNN
5. Applications in computational psychology
6. Conclusion

# Functional Network

- Neural networks have a close relationship between their **structures and functions**
- We propose the concept of the **functional network** of neural networks
- By exploring the topological structure of these functional networks, we can gain a better understanding of the connection between their structures and functions



**Fig. 1.** Flow chart of constructing the functional network for a given deep neural network.



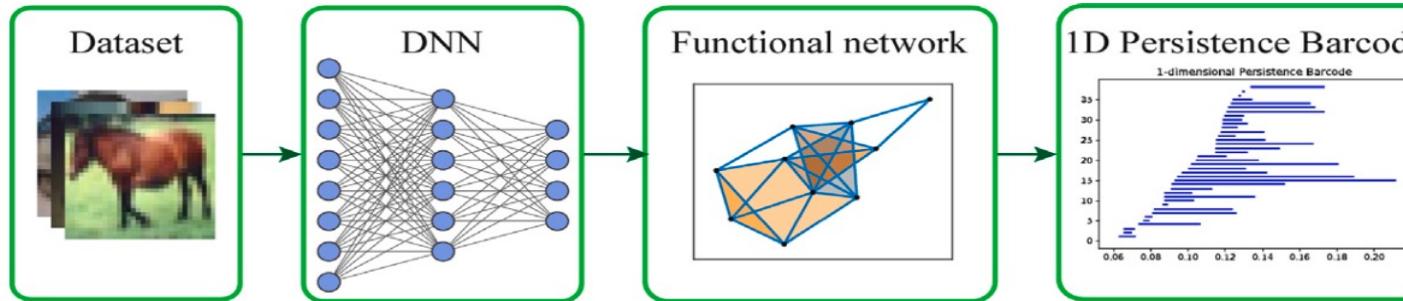
MNIST [300, 300]

Influence of different regularizations

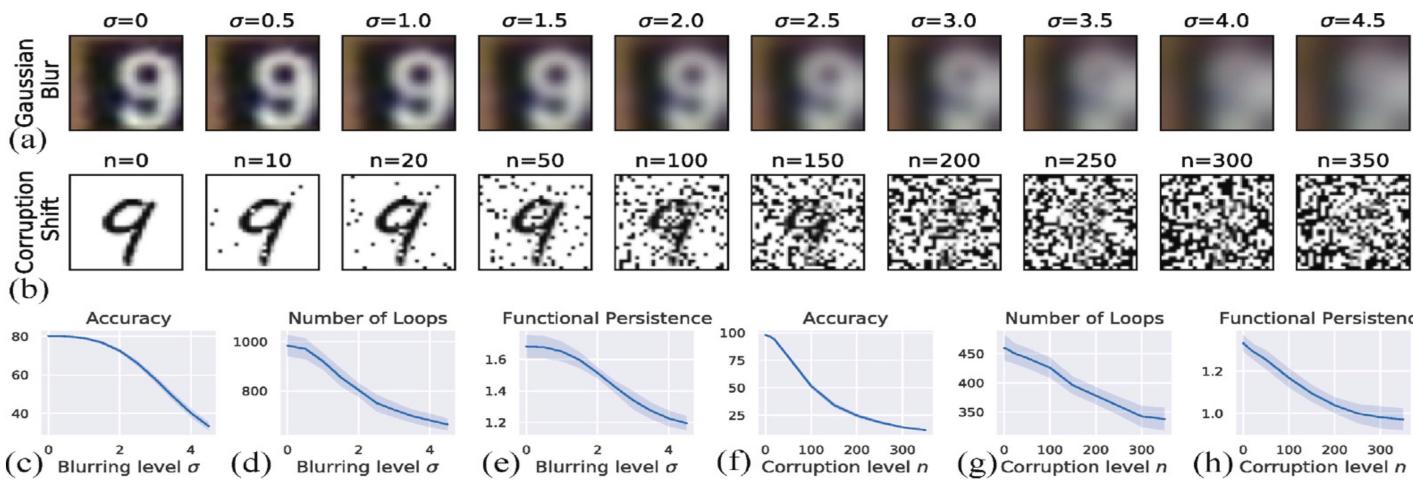
Ben Zhang, Zhetong Dong, Junsong Zhang, Hongwei Lin. Functional network: A novel framework for interpretability of deep neural networks. Neurocomputing 519: 94-103 (2023)

# Functional Loops

- The relationship between functional organizations and network performance is explored using algebraic topology
- The **functional loops** reveal **functional interaction patterns** of multiple neurons in DNNs.



- The functional network is constructed as a simplicial complex  $K(M, S)$  by computing the functional similarities between neurons
- The **1-dimensional persistent barcode** is calculated to measure the functional complexity of the DNN



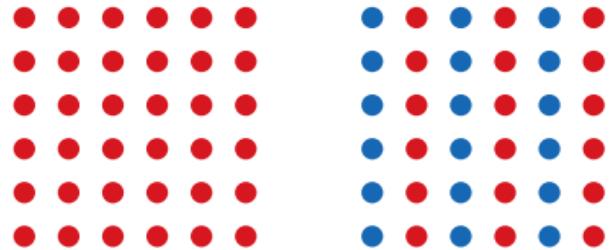
- The number of loops, functional persistence, and accuracy decrease as the dataset shifts
- Therefore, the **number of loops may reflect the number of features extracted by models** from unseen samples

Ben Zhang, Hongwei Lin. Functional loops: Monitoring functional organization of deep neural networks using algebraic topology. Neural Networks 174: 106239 (2024)

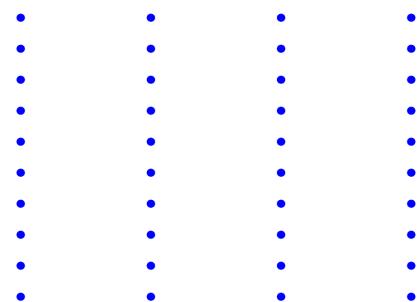


1. Introduction to computational topology
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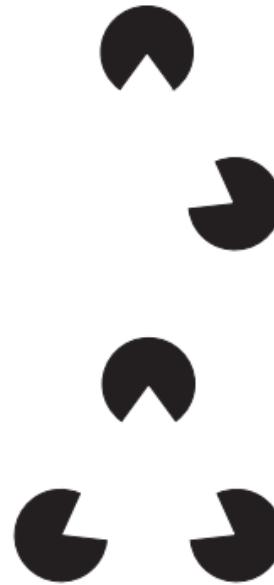
- Computational psychology studies computational models of psychological rules
- The **Gestalt theory** is a classic theory in cognitive psychology used to explain the role of visual perception
- We established the **Gestalt computational model** using persistent homology theory in computational topology
- Gestalt theory includes five core principles: similarity, proximity, closure, good continuation, and pragnanz



Similarity



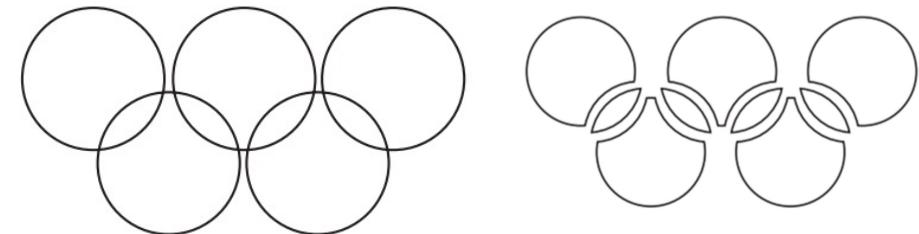
Proximity



Closure



Good continuation



Pragnanz

# Gestalt Computational Model-Step 1,2,3

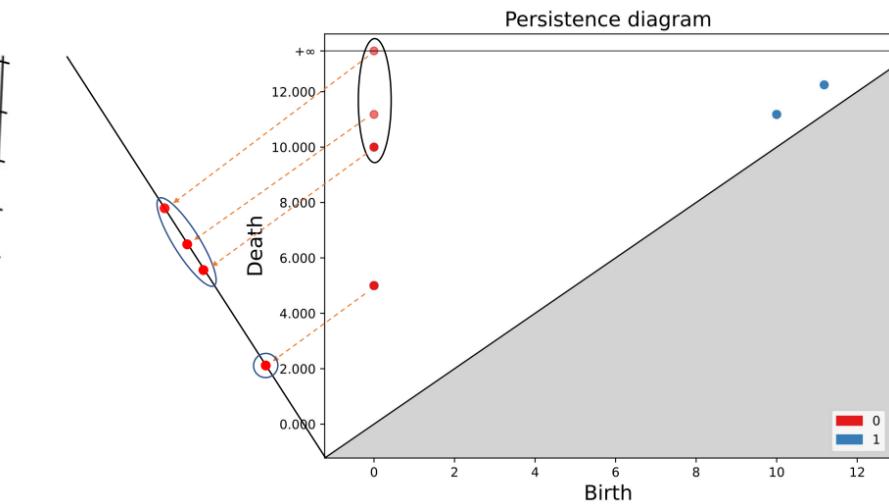
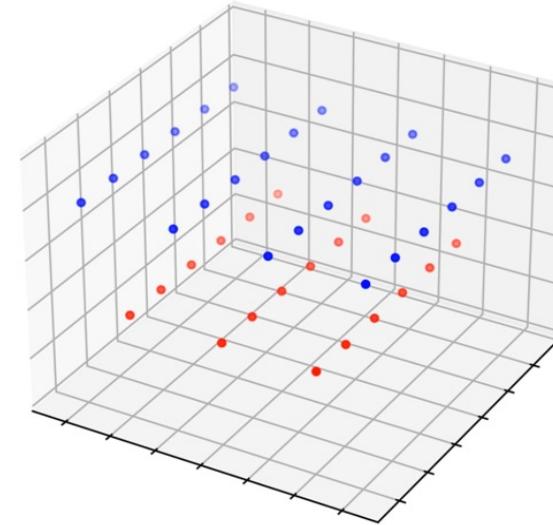
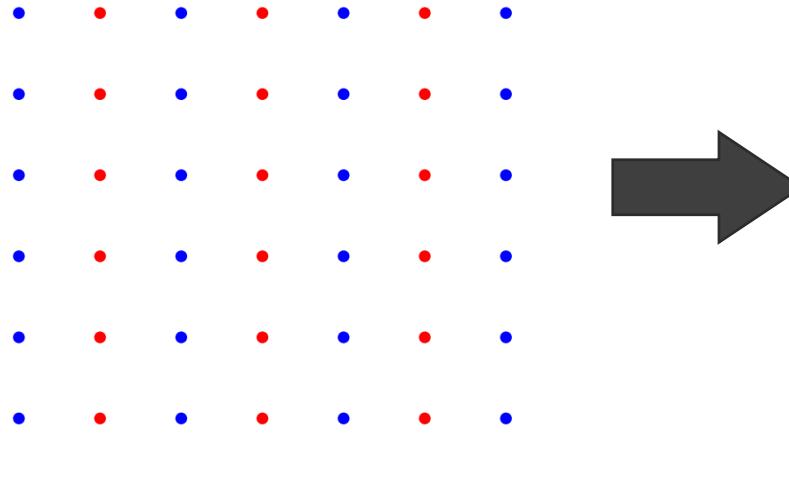
1. **Set additional coordinates:** According to the attributes of points set that affect the visual perception, assign an additional z-coordinate,

$$\{Q_i = (x_i, y_i, z_{i,1}, z_{i,2}, \dots, z_{i,m}), i = 1, 2, \dots, n\}$$

2. **Calculate PD:** Construct VR filtration and calculate the corresponding PD

3. **Cluster the points on the PD:**

- Project the points in the PD onto the line  $y = -x$  and cluster them into two categories
- The category far from the origin is the important feature points



# Gestalt Computational Model-Step 4, 5

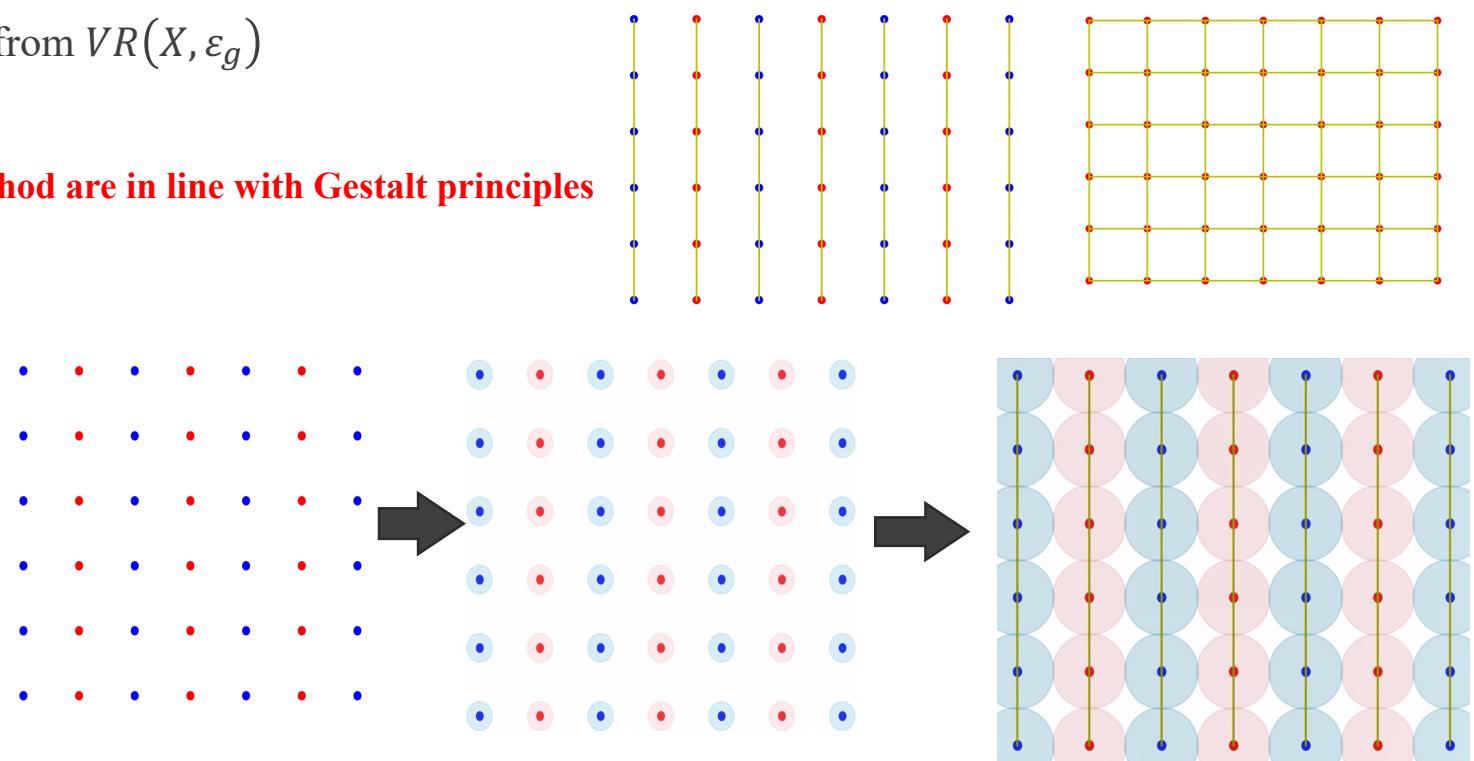
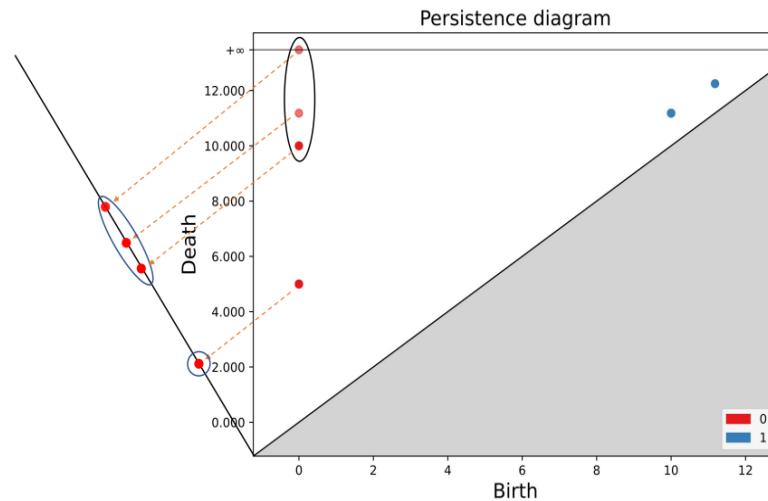
4. **Determine the threshold:** Determine a suitable threshold  $\varepsilon_g$  so that important topological features have appeared and not disappeared

- For 0-PD, select  $\varepsilon_g$  as the maximum value of the extinction values of all noise points
- For 1-PD, select  $\varepsilon_g$  as the maximum value of the birth values of all important points

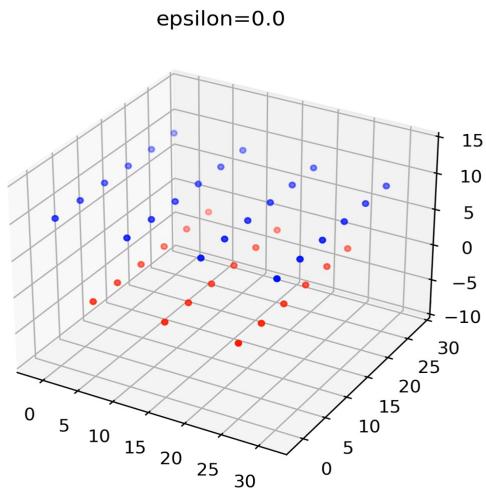
5. Reconstruction of visual perception results:

- Reconstruct the results of visual perception from  $VR(X, \varepsilon_g)$

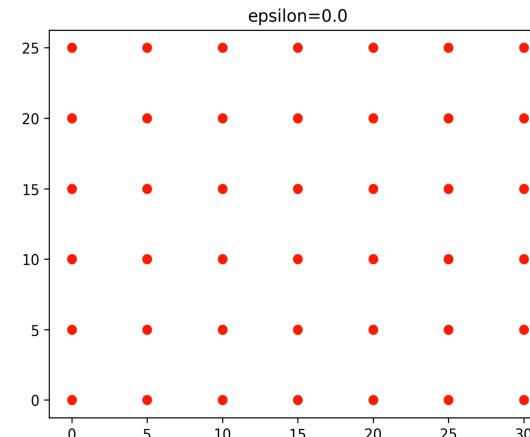
The visual perception results reconstructed by the above method are in line with Gestalt principles



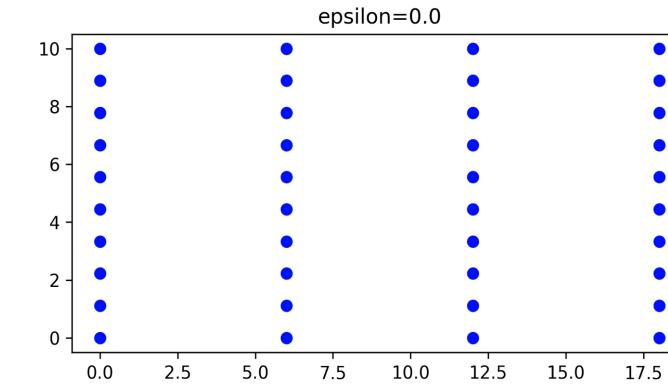
# Gestalt Computational Model: Results



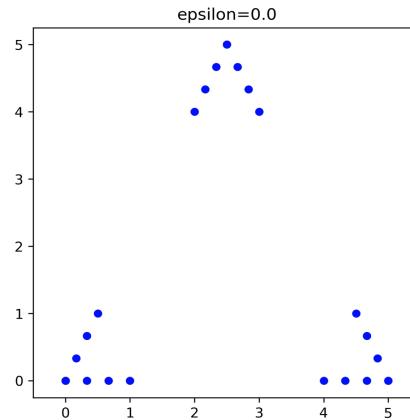
Similarity: 3D



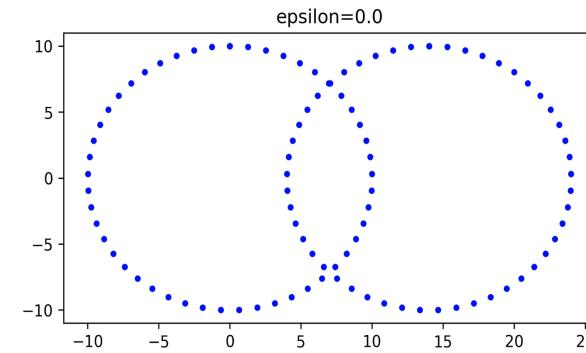
Similarity: 2D



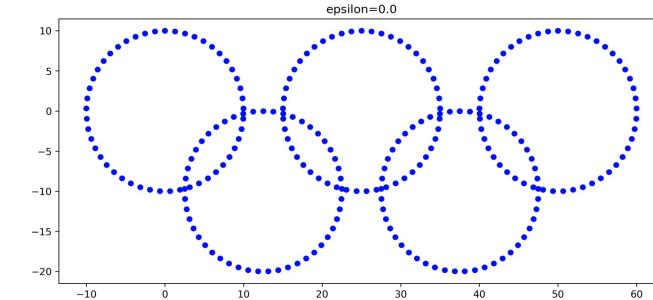
Proximity



Closure



Good continuation

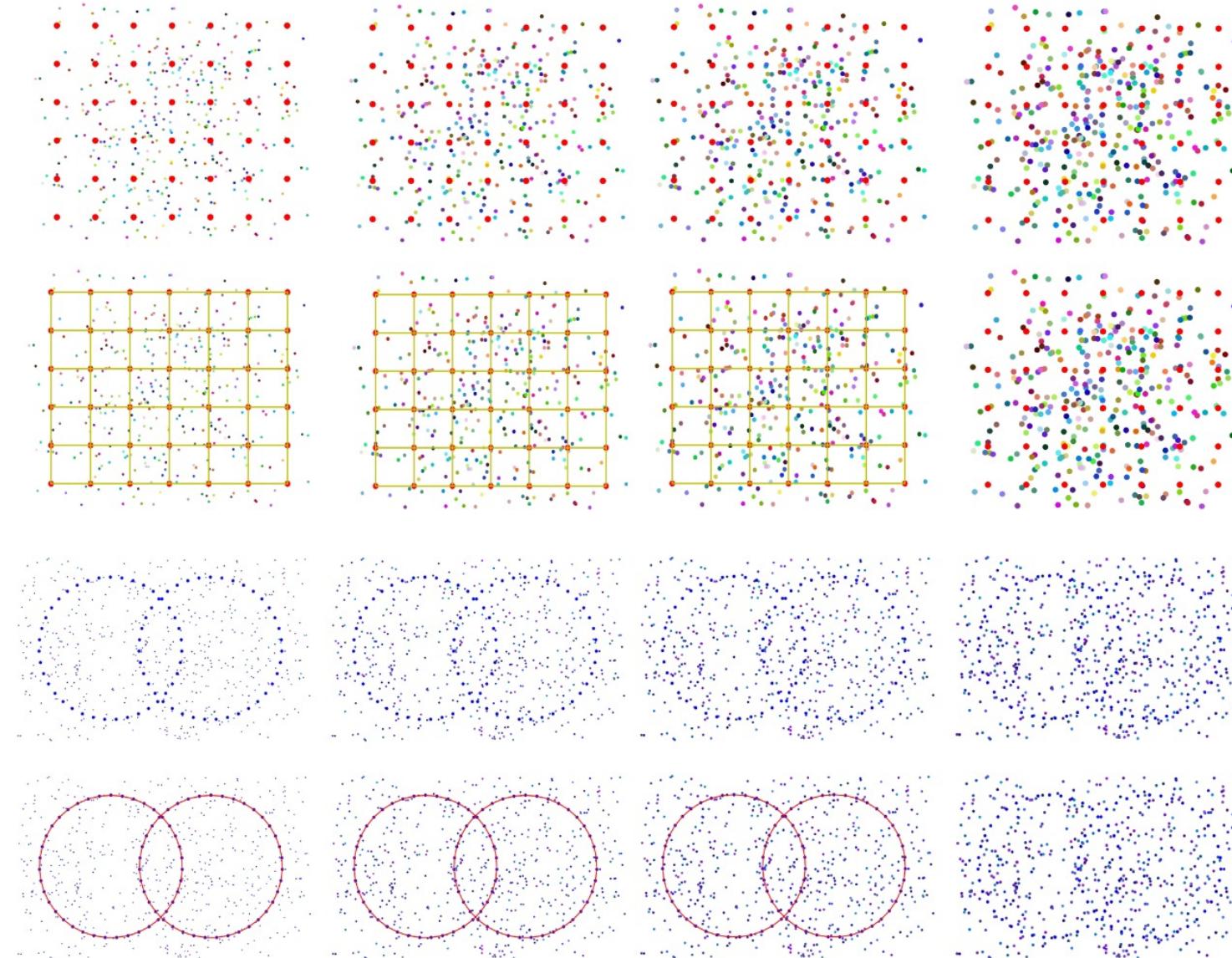


Pragnanz

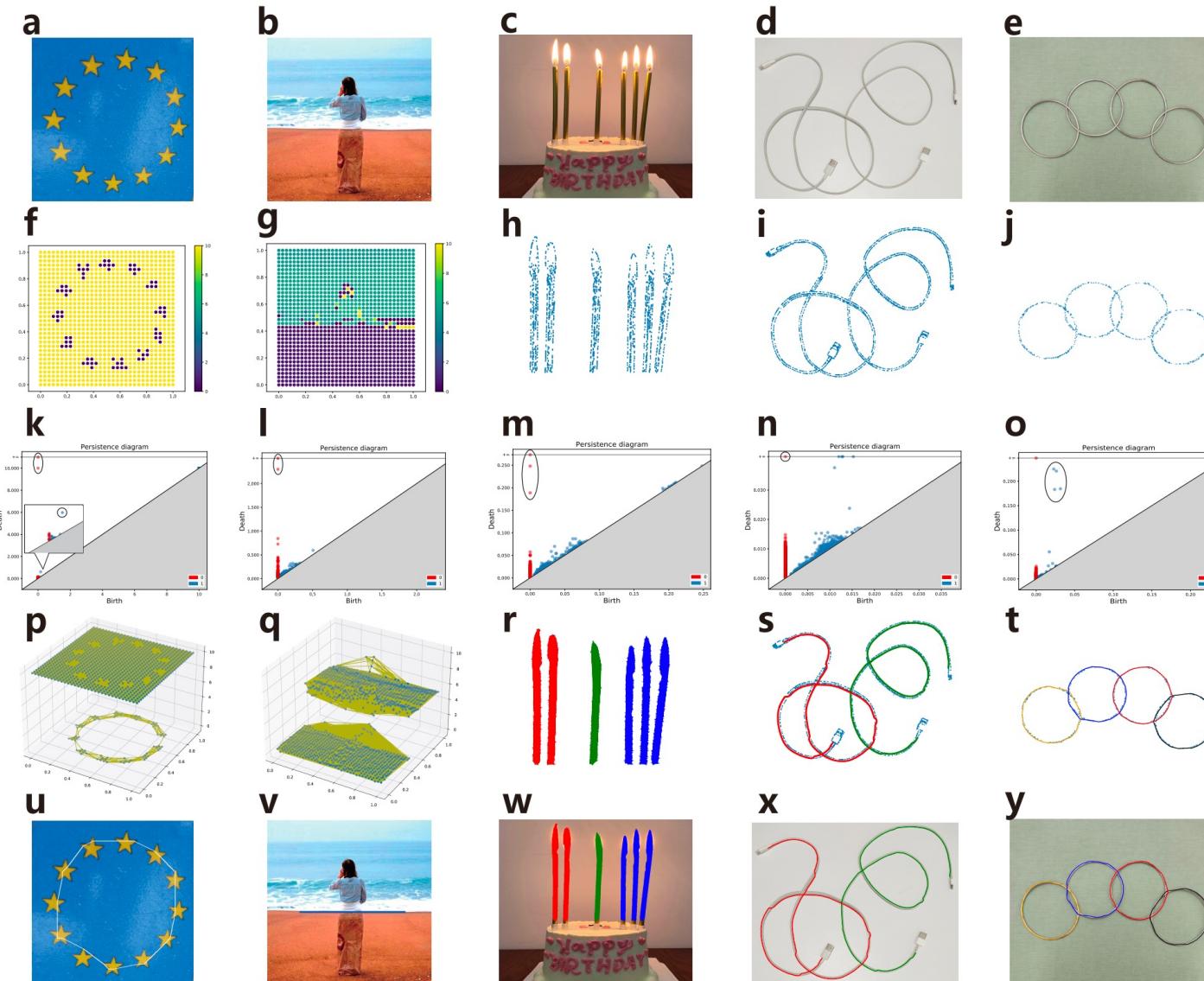
# Influence of noise



浙江大学  
ZHEJIANG UNIVERSITY



# Computational Results on Real Images





- 1. Introduction to computational topology**
- 2. Applications in geometric design**
- 3. Porous retrieval, design, and printing**
- 4. Interpretability of DNN**
- 5. Applications in computational psychology**
- 6. Conclusion**

# Conclusion



- Computational Geometry → Computational Topology
- Computer Aided Geometric Design → Computer Aided Topological Design
- 计算机辅助拓扑设计：以持续同调为主要理论工具，系统解决几何设计和几何处理中的拓扑问题

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## 计算机辅助拓扑设计

——持续同调在几何设计和处理中的应用

董哲同<sup>1,2</sup>, 蔺宏伟<sup>1,2</sup>

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2. 浙江大学CAD&CG国家重点实验室, 浙江 杭州 310058)

**摘要** 持续同调是一种计算不同尺度拓扑特征的有效方法。其从一簇向后包含的单纯复形序列中提取出拓扑特征的出现和消失时刻，并使用拓扑特征的“生命周期”来量化地衡量该特征的几何尺度和重要程度。拓扑特征的提取与应用在几何设计中扮演着重要角色，催生出了一些基于持续同调的几何设计研究。从持续同调特征的提取与基于持续同调的建模和优化两方面进行综述，在持续同调特征的提取方面，介绍了从点云和三角网格数据中提取拓扑特征的不同方法，总结了拓扑特征在部分几何设计问题中的应用路径。在建模和优化方面，综述了基于拓扑变换的单纯复形重建方法、拓扑可感知的曲面重建方法与基于持续同调的拓扑去噪和优化方法。

**关键词** 持续同调；拓扑特征提取；形状重建；去噪与优化；几何设计；拓扑设计

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文章编号: 2095-302X(2022)06-0957-10

## Computer aided topological design

——survey on geometric design and processing based on persistent homology

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(1. School of Mathematical Sciences, Zhejiang University, Hangzhou Zhejiang 310027, China;  
2. State Key Laboratory of CAD&CG, Zhejiang University, Hangzhou Zhejiang 310058, China)

已发表或录用近30篇学术论文：

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- 深度神经网络的可解释性
- 股票数据处理

谢 谢！



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