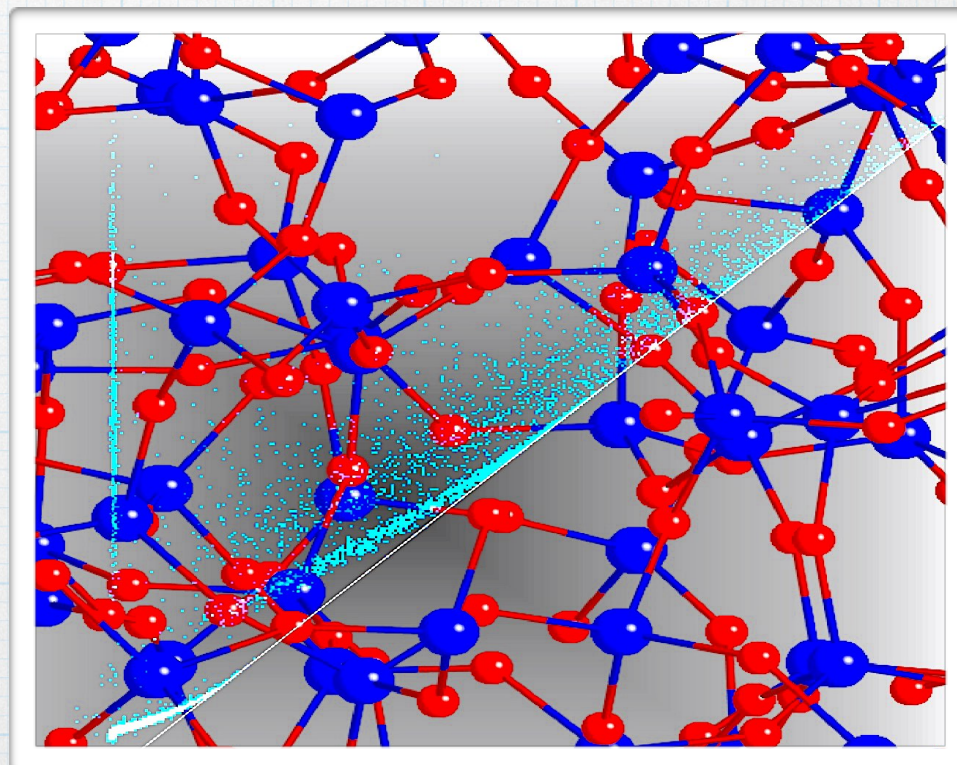


パーシステントホモロジーと機械学習

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東北大学
原子分子材料科学高等研究機構
WPI-AIMR

JST CREST
SIP革新的構造材料
JST イノベーションハブMI²I



位相的データ解析

位相的データ解析 (Topological Data Analysis, TDA)

今世紀に数学者が開発したデータ解析手法

Data Has Shape, Shape Has Meaning, Meaning Drives Value

Gunnar Carlsson's Gr. (math. Stanford, AYASDI)

- (ビッグ) データ解析, ソーシャルネットワーク, 医療, 金融 etc

Robert Ghrist's Gr. (math. UPenn)

- 情報ネットワーク, センサーネットワーク

Konstantin Mischaikow's Gr. (math. Rutgers)

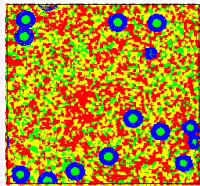
- 流体解析, データ時系列解析

東北大AIMR, CREST, SIP, MI²I

- 材料科学 (ガラス, 粉体, 高分子, 金属, 蓄電池 etc)

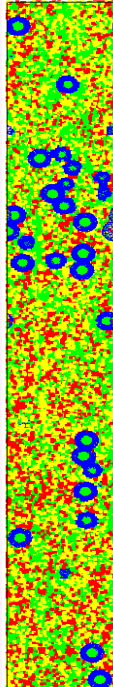
位相的データ解析の材料科学への応用

高分子



AFM
image

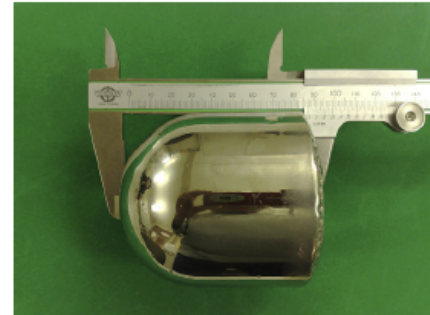
expansion



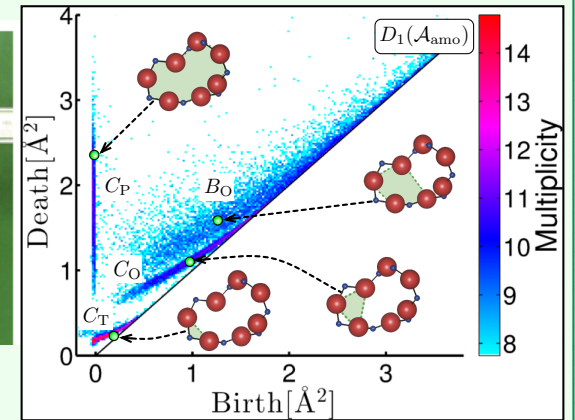
アモルファス



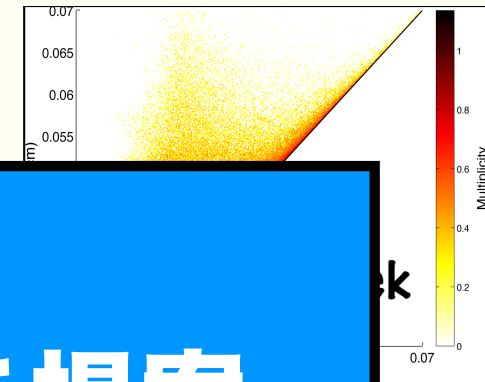
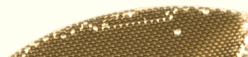
silica



metallic glass

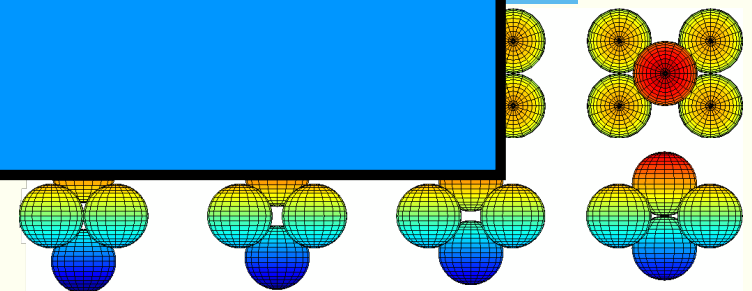


粉体



無秩序系の記述言語を提案

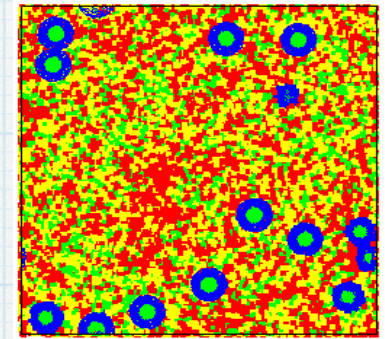
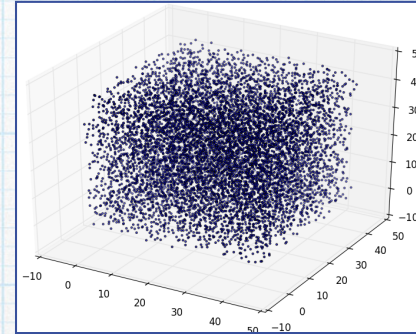
JOINT WITH ANU
(M.Saadatfar etc)



Shape of Data

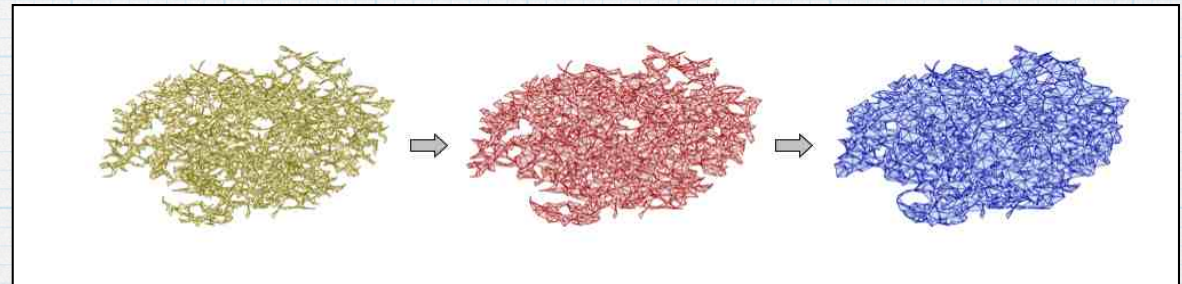
Data

point cloud, atomic configuration, digital image, sensor network



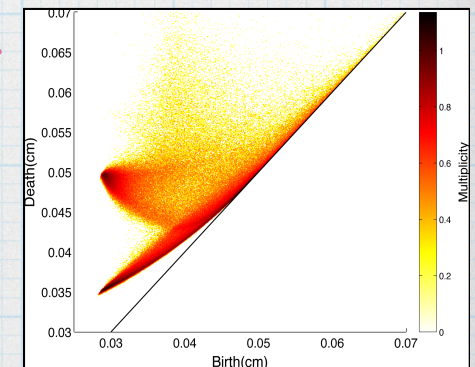
Geometric Model

Polyhedron (Alpha, Čech, cell complex)



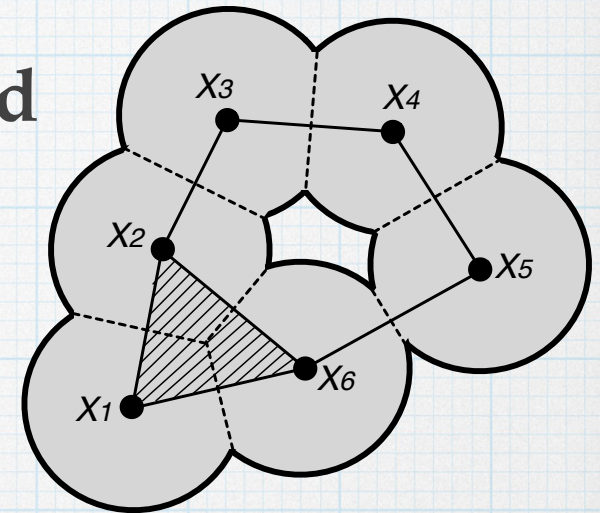
Algebraic Description

Persistent homology, sheaf cohomology, category

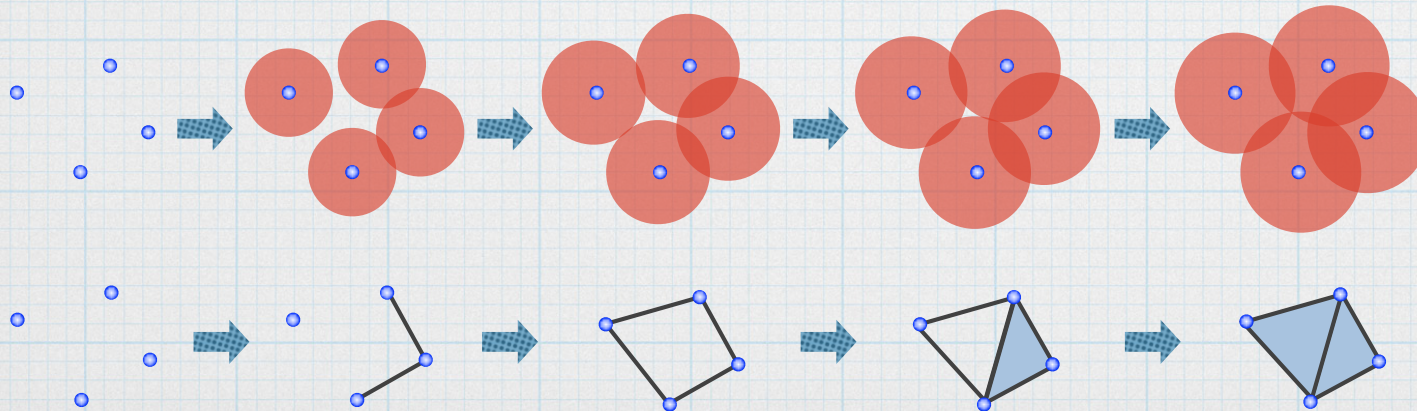


Alpha shape

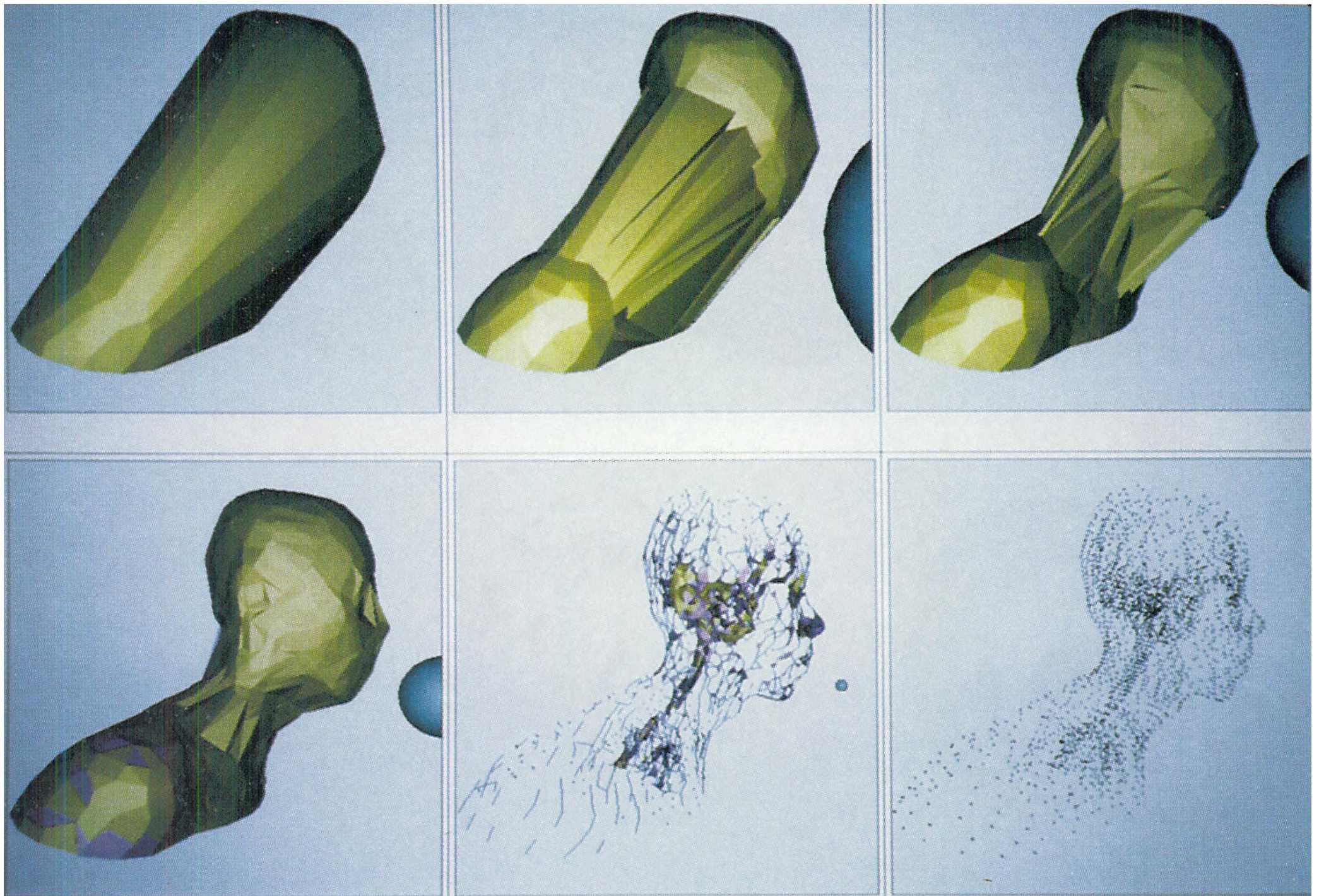
- $X = \{x_i \in \mathbf{R}^m \mid i = 1, \dots, n\}$: point cloud
- $\mathbf{R}^m = \cup_i V_i$: Voronoi decomp.
- $\cup_i B_i(r) = \cup_i (B_i(r) \cap V_i)$
- Alpha shape $\mathcal{A}(X, r)$: dual of $\{B_i(r) \cap V_i \mid i = 1, \dots, n\}$
(simplicial complex)
- Nerve theorem: $\cup_i B_i(r) \simeq \mathcal{A}(X, r)$
- $\mathcal{A}(X, r) \subset \mathcal{A}(X, s)$ for $r < s$



easier to analyze by computers

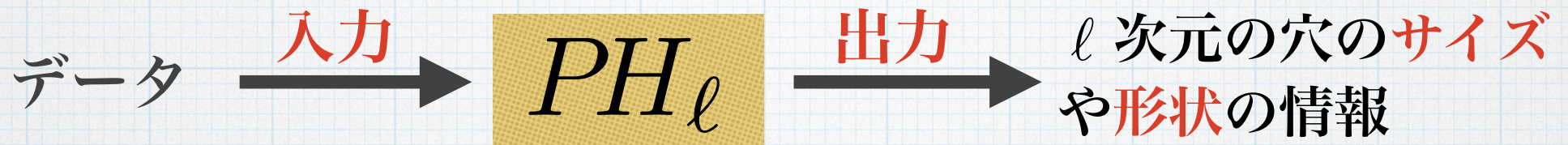


filtration:
changing resolution

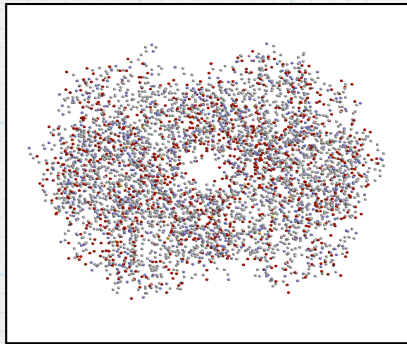


(ref. Edelsbrunner, Mücke)

パーシステントホモロジー

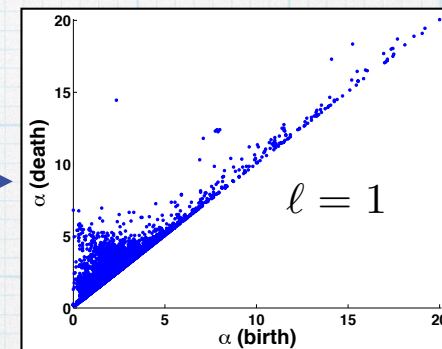


ヘモグロビン



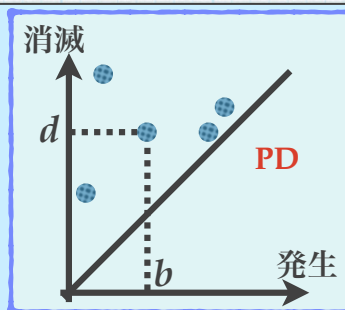
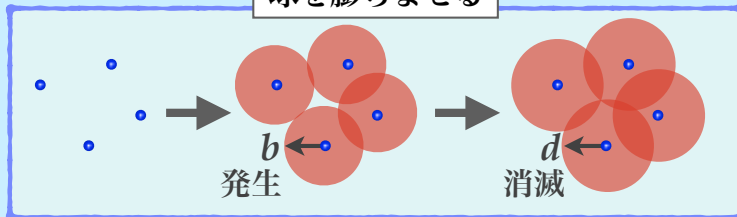
パーシステントホモロジー

パーシステント図(PD)



PD構成法

球を膨らませる

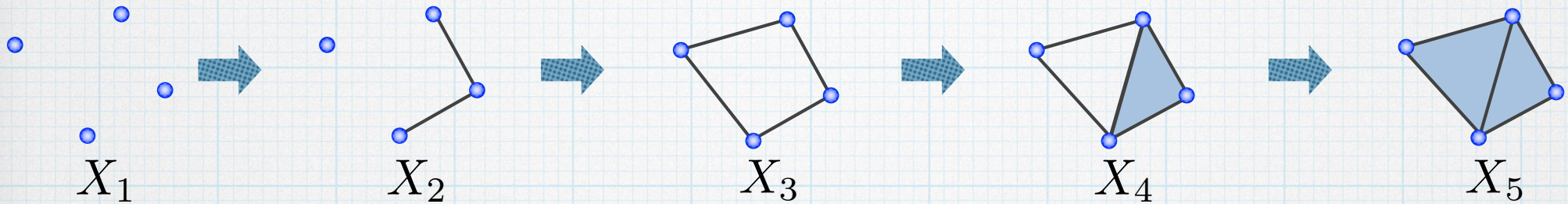


- PDの各点はデータ内の穴
- 発生軸は穴の発生パラメータ
- 消滅軸は穴の消滅パラメータ
- 対角線付近の点はノイズ
- 対角線から離れた点はロバスト

- 数学者Edelsbrunner, Carlsson等によって開発 (2003, 2005年)
- 穴のサイズ, 形状, 階層構造などを扱える
- 数学的進化と同時に, 現在急速に諸科学への応用が進められている

Persistent Homology

$$PH_1(\mathcal{X}) \simeq I[3, 4]$$



• **filtration** $\mathcal{X} : X_1 \subset X_2 \subset \cdots \subset X_n$

• **persistent homology**

representations on A_n

$$H_\ell(\mathcal{X}) : H_\ell(X_1) \rightarrow H_\ell(X_2) \rightarrow \cdots \rightarrow H_\ell(X_n)$$



• **interval decomposition (Gabriel's Theorem)**

$$H_\ell(\mathcal{X}) \simeq \bigoplus_{i=1}^s I[b_i, d_i]$$

generator

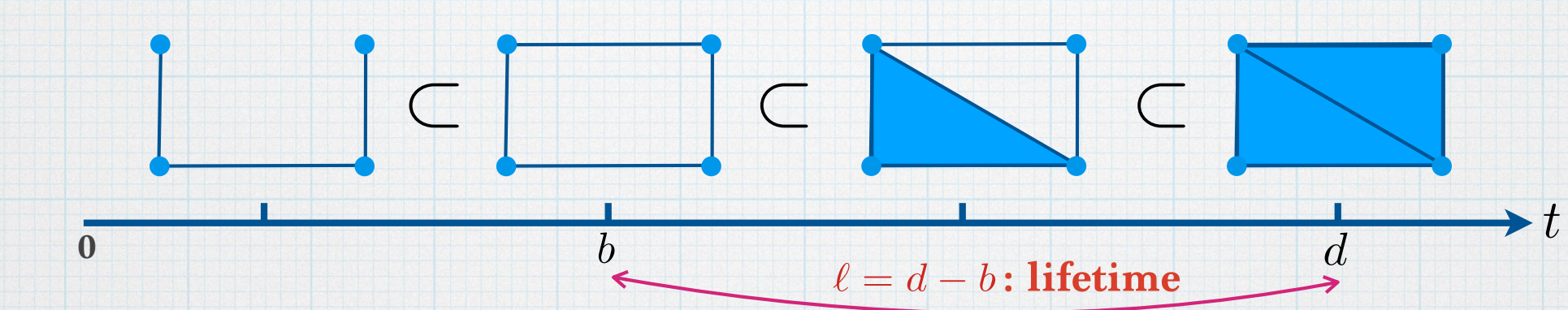
$$I[b, d] : 0 \rightarrow \cdots \rightarrow 0 \rightarrow K \rightarrow \cdots \rightarrow K \rightarrow 0 \rightarrow \cdots \rightarrow 0$$

at X_b at X_d

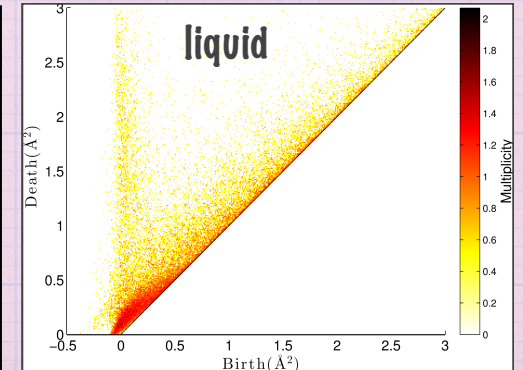
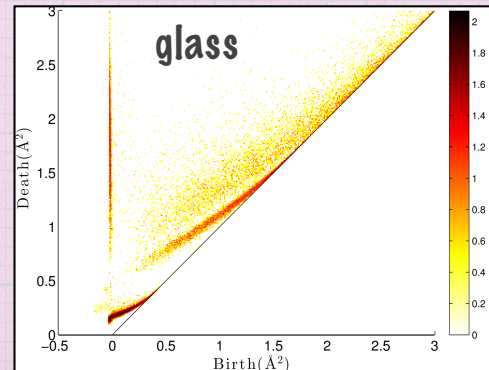
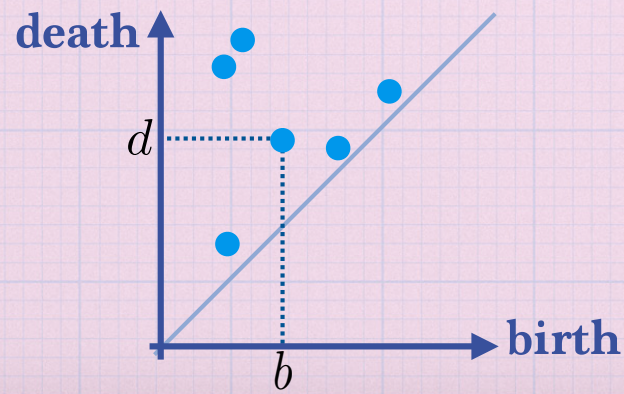
Persistence Diagram

Interval decomp: $H_k(\mathcal{X}) \simeq \bigoplus_{i=1}^p I[b_i, d_i]$

- $I[b, d]$ represents appearance and disappearance of a topological feature at $t = b, d$ in $\mathcal{X} = \{X(t)\}_t$

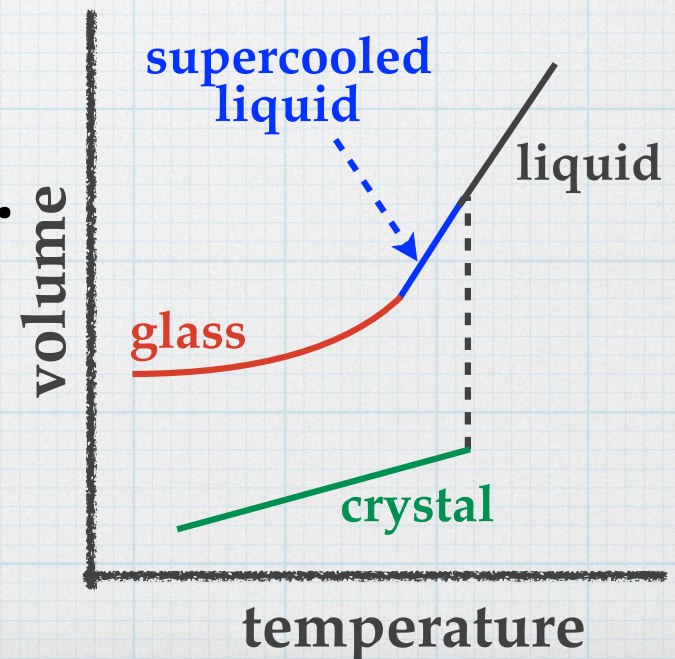


- $D_k(\mathcal{X}) = \{(b_i, d_i) \in \mathbb{R}_{\geq 0}^2 : i = 1, \dots, p\}$: **persistence diagram**

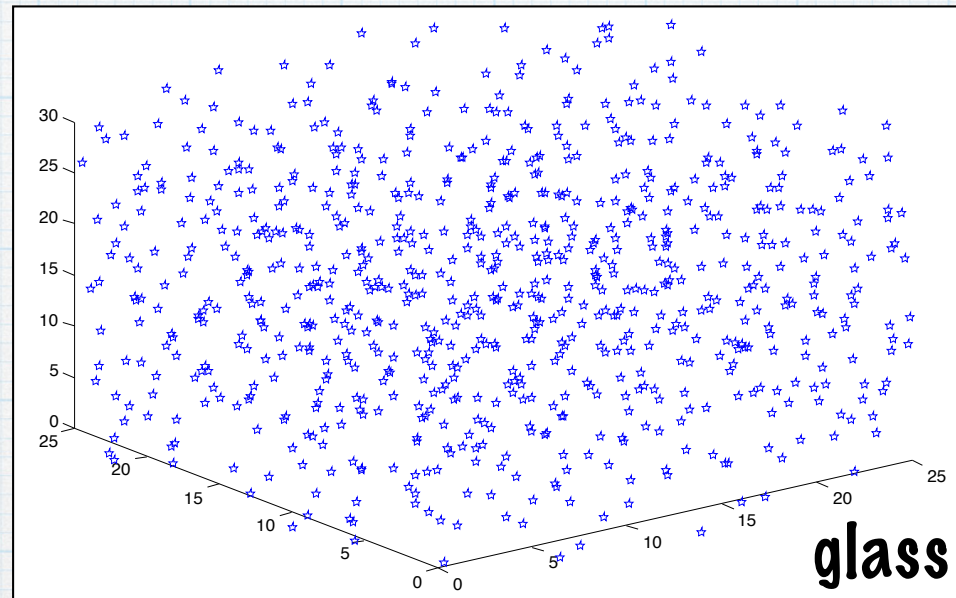


What is glass?

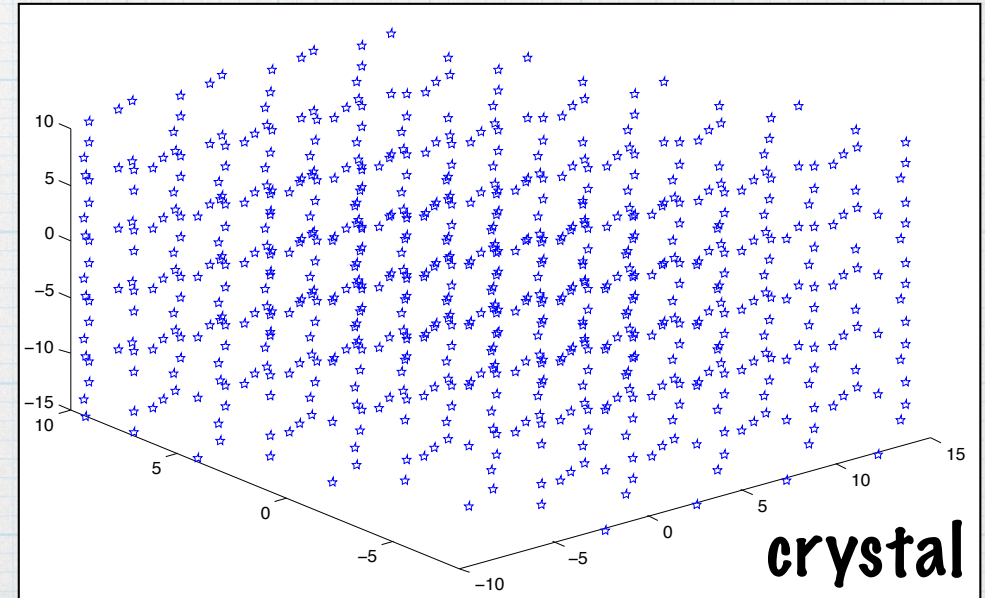
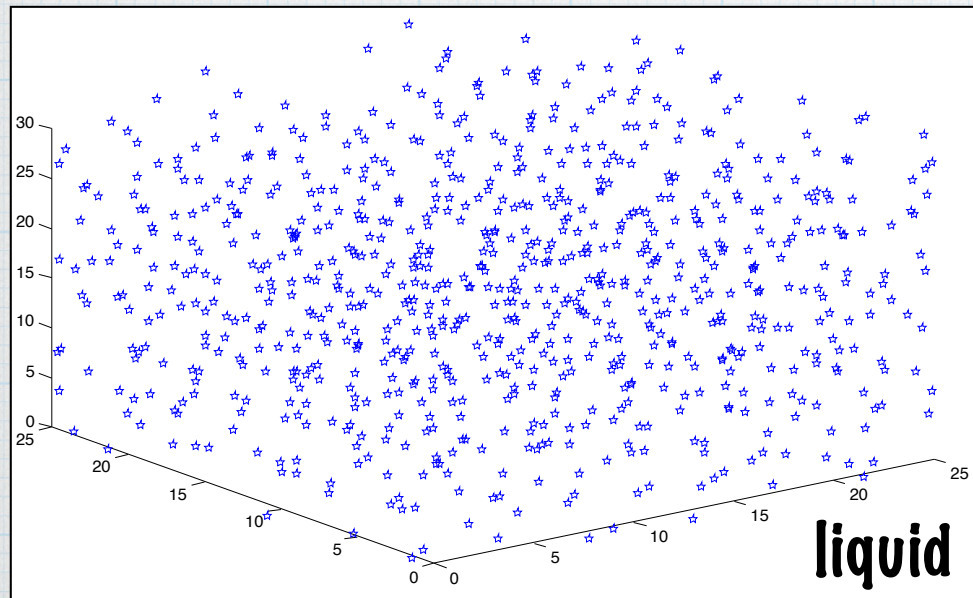
- * Not yet fully answered to “what is glass?”
- * Not liquid, not solid, but something in-between
- * Atomic configuration looks random, but sufficient cohesion to maintain rigidity
- * Further geometric understandings of atomic configurations are required
- * Solar Energy Glass, DVD, BD, etc.



Atomic configurations of silica (SiO_2)

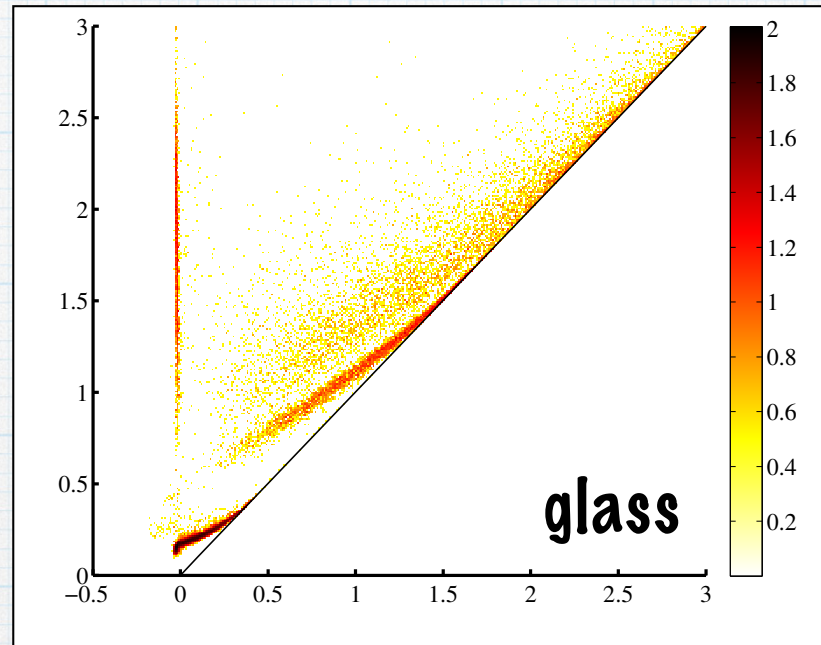


obtained by
molecular
dynamics
(MD)
simulation

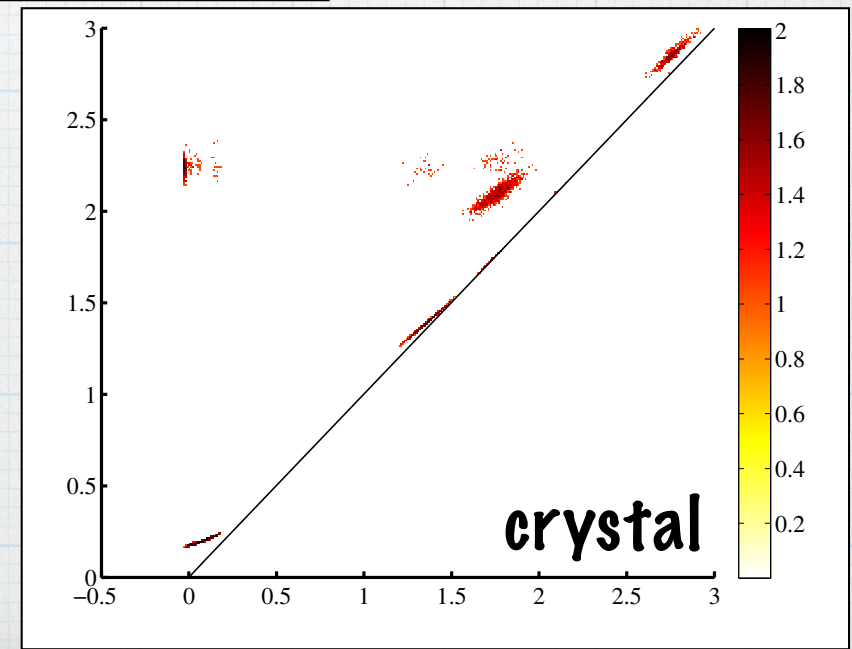
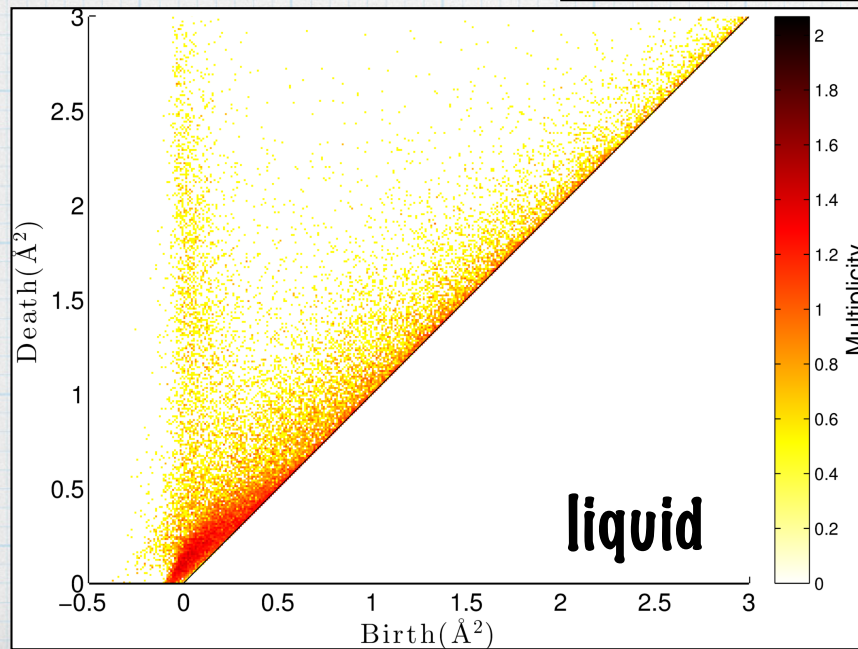


1 dim persistence diagrams of silica

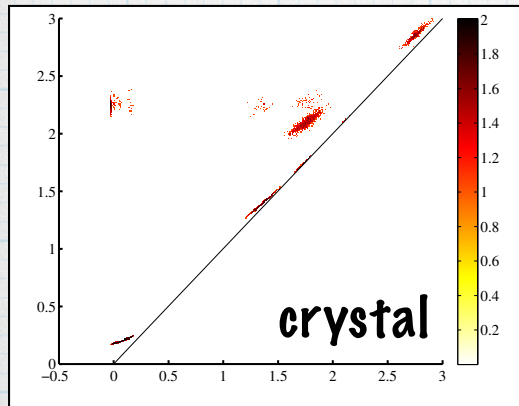
All PD computations are performed by HomCloud



PD₁ of alpha filtrations

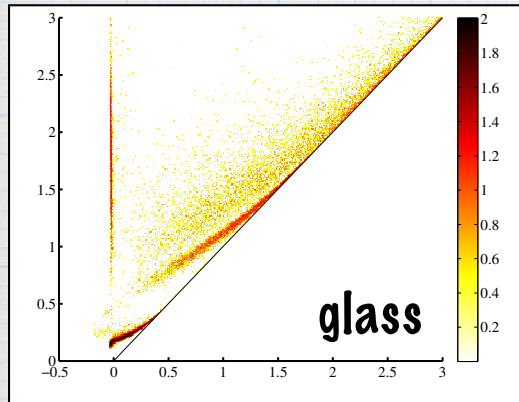


Support dim and order parameter



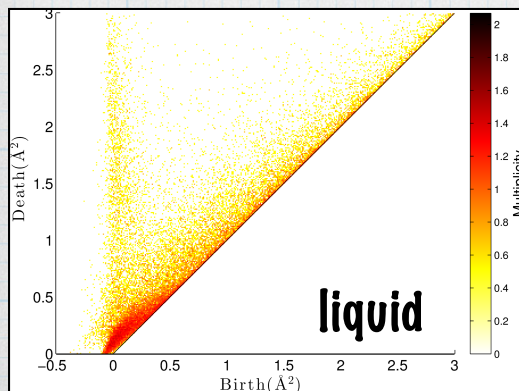
0-dim support results from **periodic** atomic locations of crystals

transition



1-dim support (curves!) appears

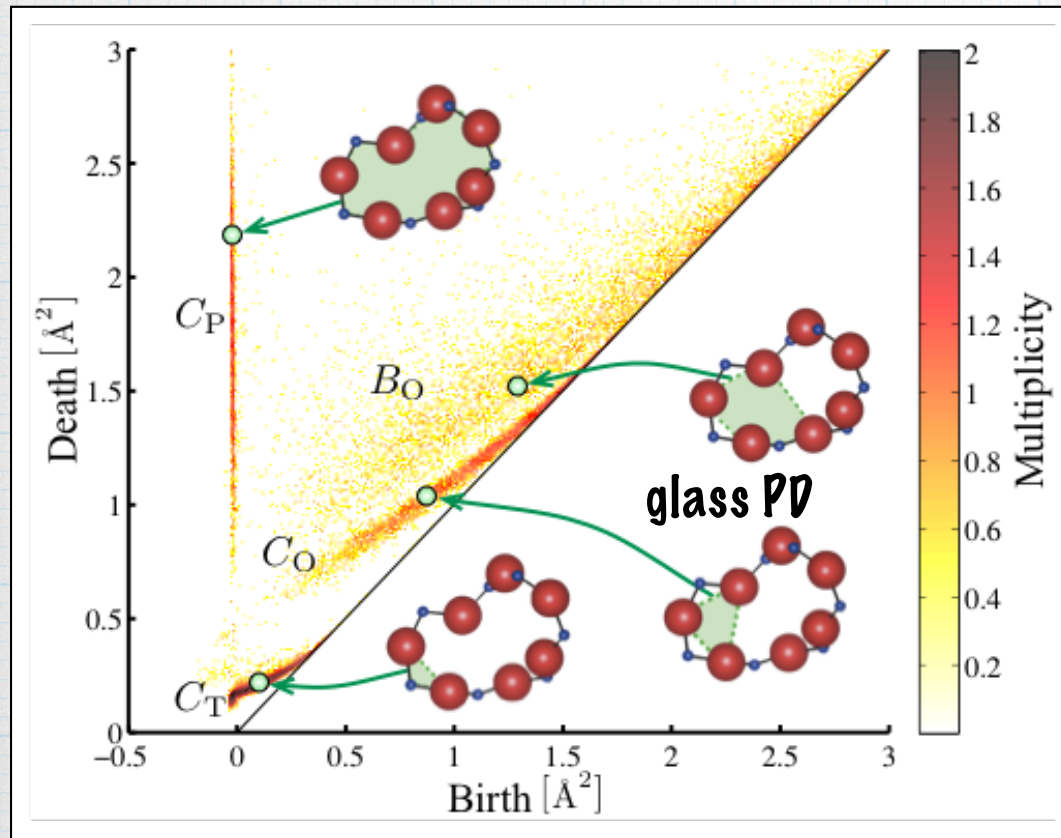
transition



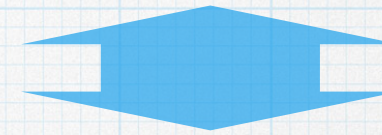
2-dim support results from **random** atomic locations of liquids

Remark : This suggests a topological approach for glass transition.

Geometric origins of curves: inverse problem

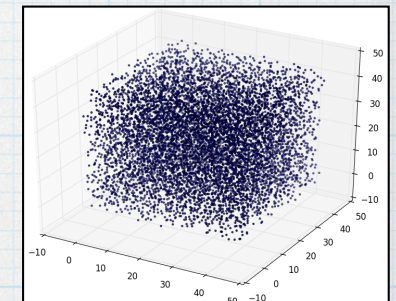


what characterizes glass structures?



what is the geometric origin of the curves?

inverse problem!

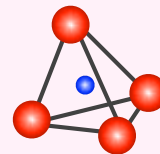


- optimal cycle
Dey, et.al, 2011, Escolar and H, 2015
- continuation
Gameiro, Obayashi, H. Physica D, 2015

hierarchical ring structure

C_P: primary rings generating the others → C_O: three oxygen rings

C_T: triangles on tetrahedra



B_O: oxygen rings (\geq four)

PD and kernel method

- * vectorization of PDs for statistics (persistence landscape, etc)
- * Reininghaus et al. proposes the persistence scale-space kernel (PSSK) on PDs based on a heat diffusion
- * Our method: **persistence weighted Gaussian kernel (PWGK)**
- * PWGK can explicitly control the effect of persistence

Let Ω be a LCH and $k : \Omega \times \Omega \rightarrow \mathbb{R}$ be a positive def. kernel

\mathcal{H}_k : reproducing kernel Hilbert space (RKHS) s.t. $f(x) = \langle f, k(\cdot, x) \rangle_{\mathcal{H}_k}$
 $M_b(\Omega)$: space of all finite signed Radon measures on Ω $\forall x \in \Omega, \forall f \in \mathcal{H}_k$

k is called **C_0 -universal** if

- $k(x, x)$ is a C_0 -function (cont. & vanishing at ∞)
- \mathcal{H}_k is dense in $C_0(\Omega)$

Thm: Kernel embedding of distributions (Fukumizu et al, 2011)

- $\Phi_k : M_b(\Omega) \rightarrow \mathcal{H}_k, \mu \mapsto \int k(\cdot, x) d\mu(x)$ is injective.
- $d_k(\mu, \nu) = \|\Phi_k(\mu) - \Phi_k(\nu)\|_{\mathcal{H}_k}$ defines a distance on $M_b(\Omega)$.

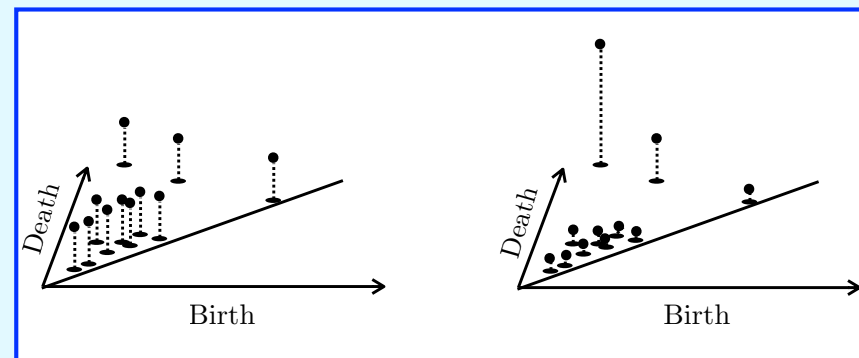
Persistence weighted Gaussian kernel (PWGK)

- weighted persistence diagram

$$\mu_D^w = \sum_{x \in D} w(x) \delta_x$$

ex) $w_{\text{arc}}(x) = \tan^{-1}(C \text{pers}(x)^p)$

C and p explicitly control the effect of persistence in statistics.



- PWGK

The Gaussian kernel $k_G(x, y) = e^{-\frac{\|x-y\|^2}{2\sigma^2}}$ is Co-universal

➡ $\Phi_{k_G} : M_b(\Delta) \rightarrow \mathcal{H}_{k_G}, \mu_D^w \mapsto \int k_G(\cdot, x) d\mu_D^w(x)$ becomes injective

$d_{k_G}^w(D, E) = \|\Phi_{k_G}(\mu_D^w) - \Phi_{k_G}(\mu_E^w)\|_{\mathcal{H}_{k_G}}$ defines a distance

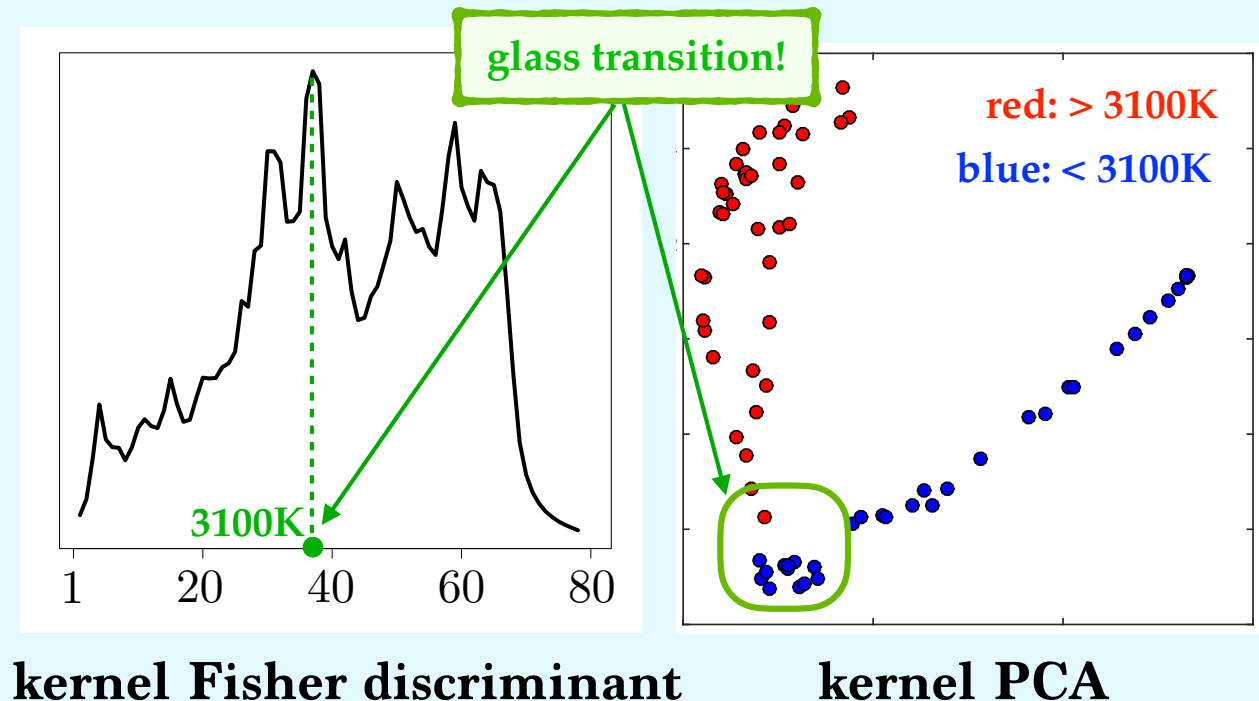
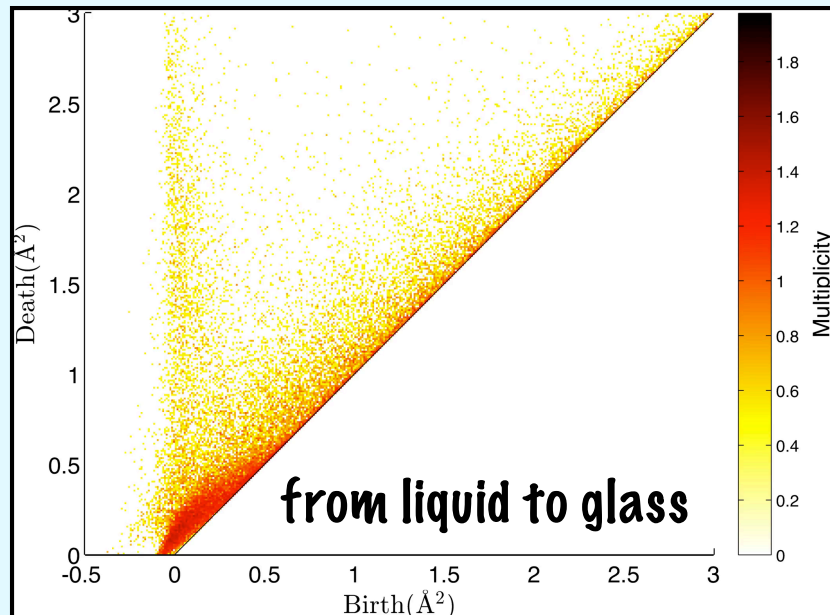
- Stability Theorem

Let $X, Y \subset \mathbb{R}^d$ be finite subsets and $p > d + 1$. Then,

$$d_{k_G}^{w_{\text{arc}}}(D(X), D(Y)) \leq L d_H(X, Y)$$

PD detects glass transition and its geometry

PWGK

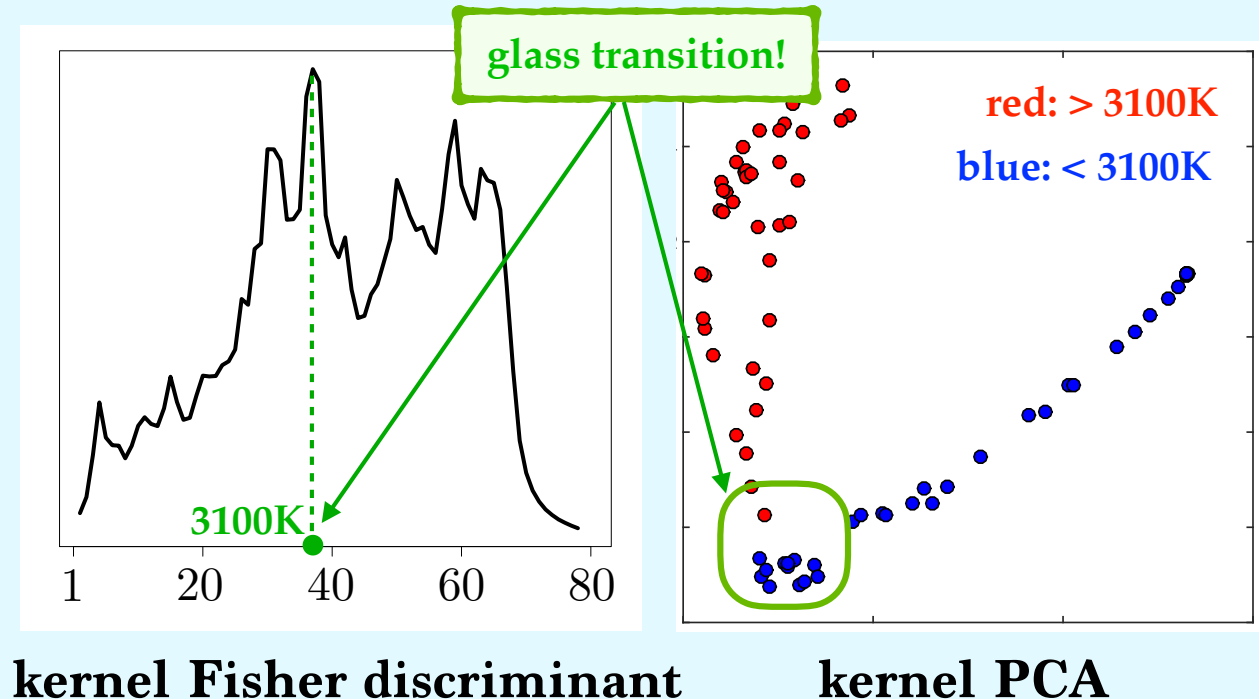
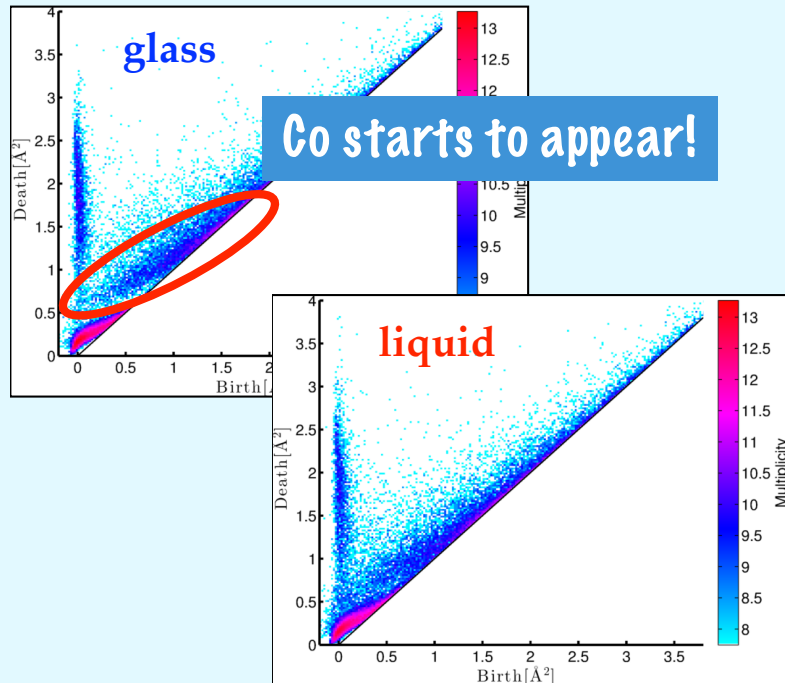


- ◆ PWGK successfully detects GT
- ◆ What is the geometric trigger to generate the phase transition?

➡ **Inverse problem!**

PD detects glass transition and its geometry

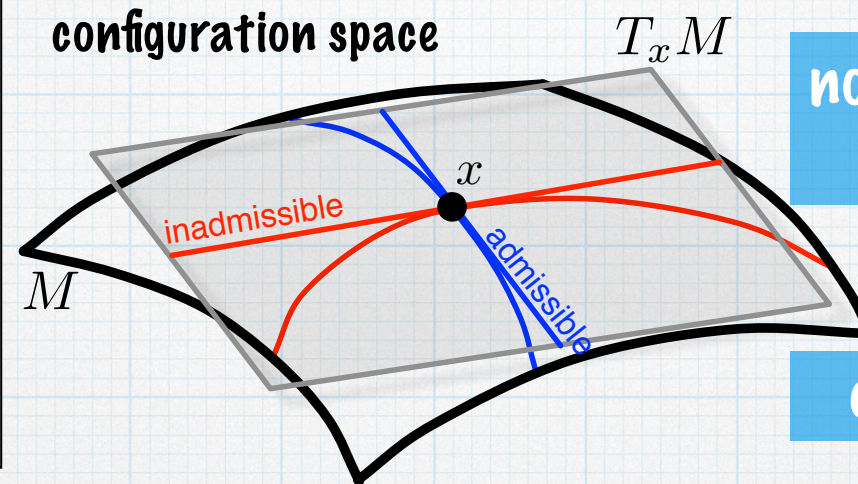
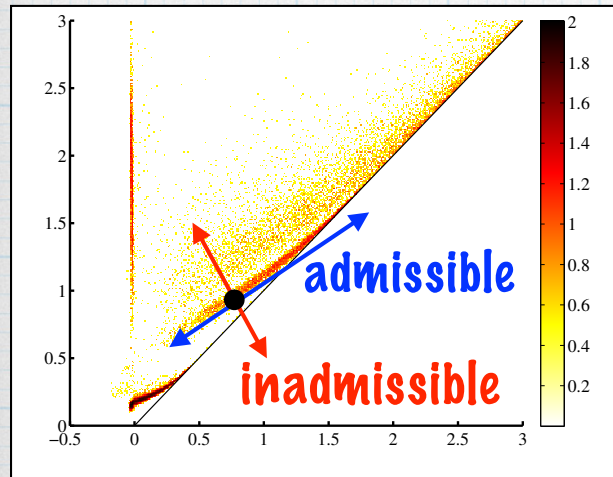
PWKG



- ◆ PWGK successfully detects GT
- ◆ What is the geometric trigger to generate the phase transition?

➡ Inverse problem!

Curves and constraints

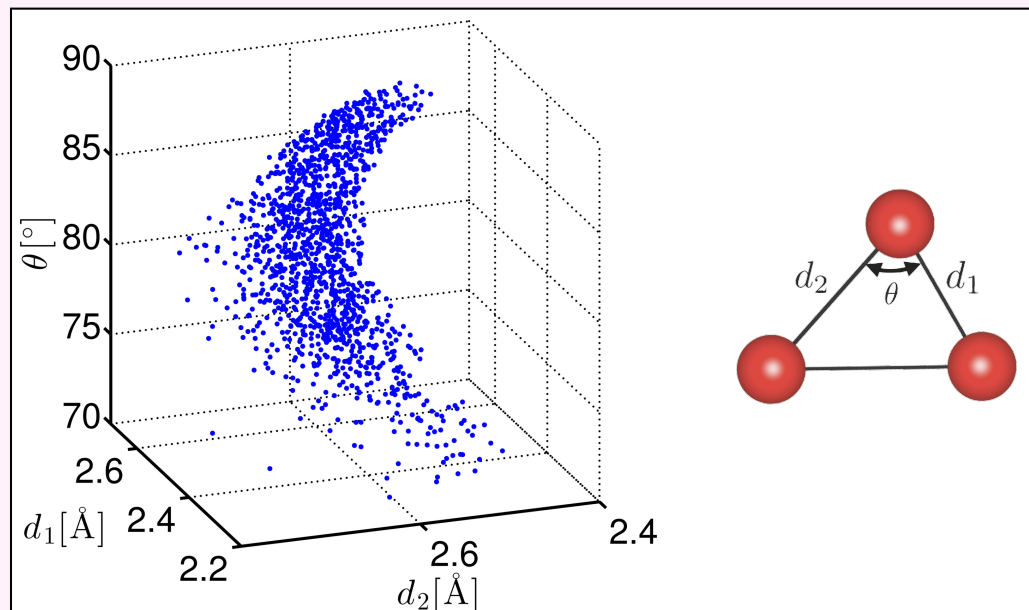


not allowed to move in the
normal direction

rigidity

constraints of the rings

O-O-O structures (MR0)



* O-O-O ring constraints are discovered

* necessary to study both distance and angle distributions simultaneously (conventional methods cannot detect)