



TDA in Engineering Applications



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Outline

- 1. Introduction
- 2. Applications

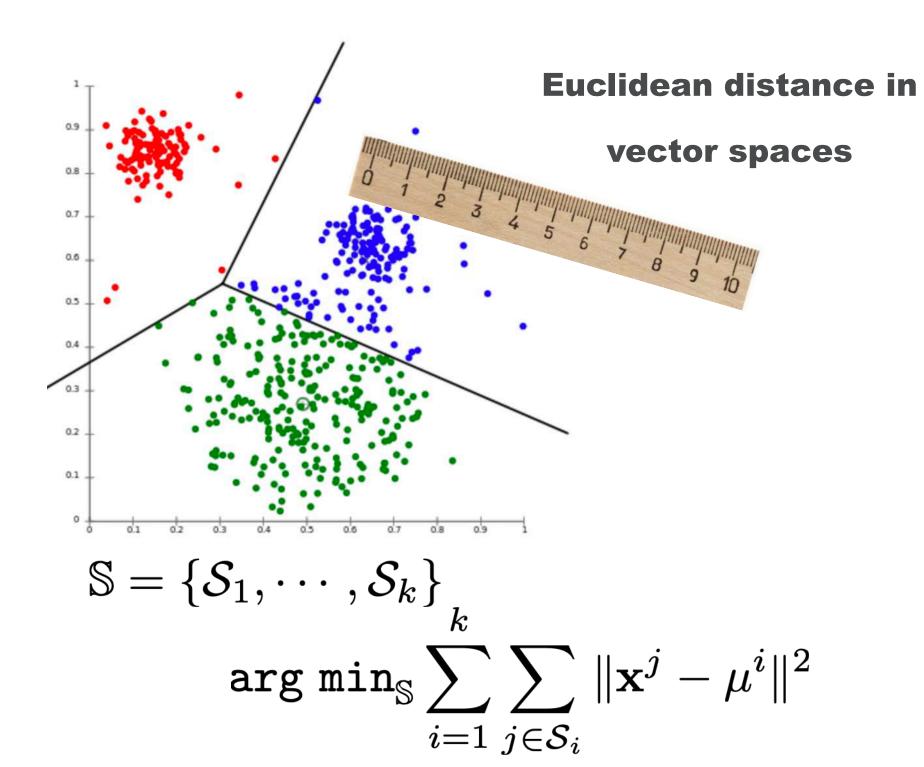
1. Introduction

2. Applications

Data in a vector space

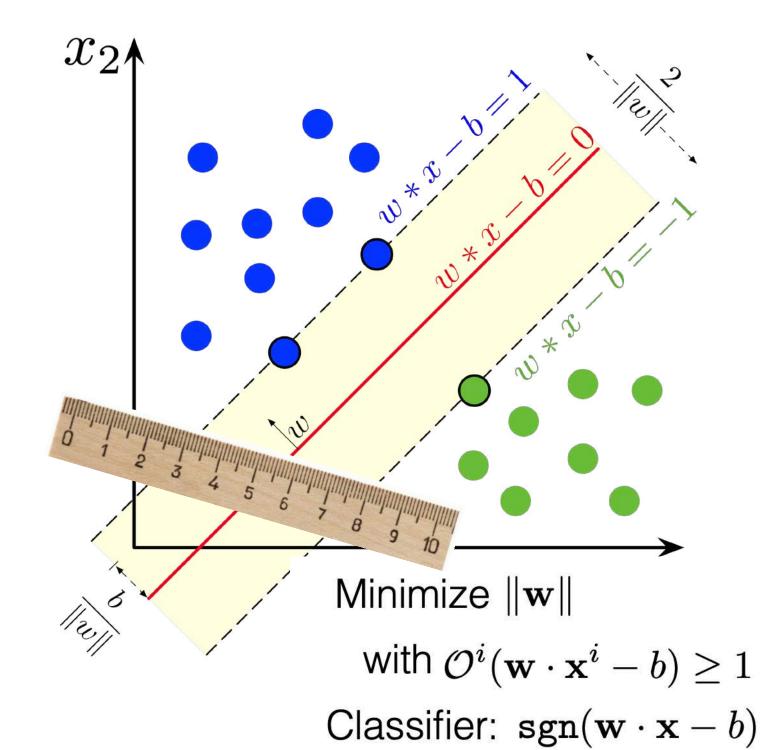
Unsupervised

clustering

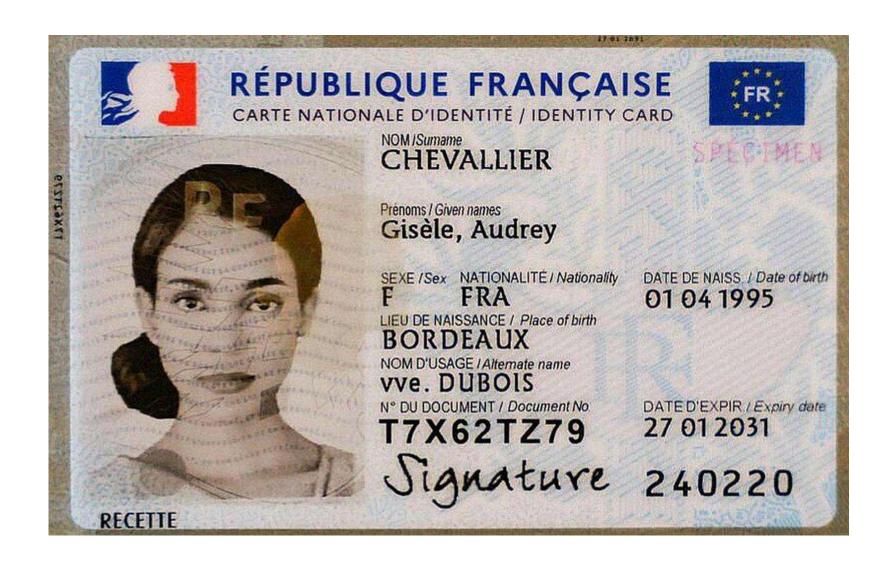


Supervised

classification



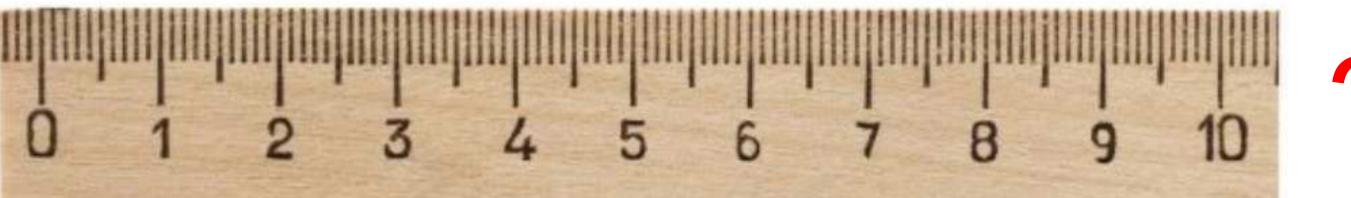
Complex Data



Data can contain multiple heterogeneous information

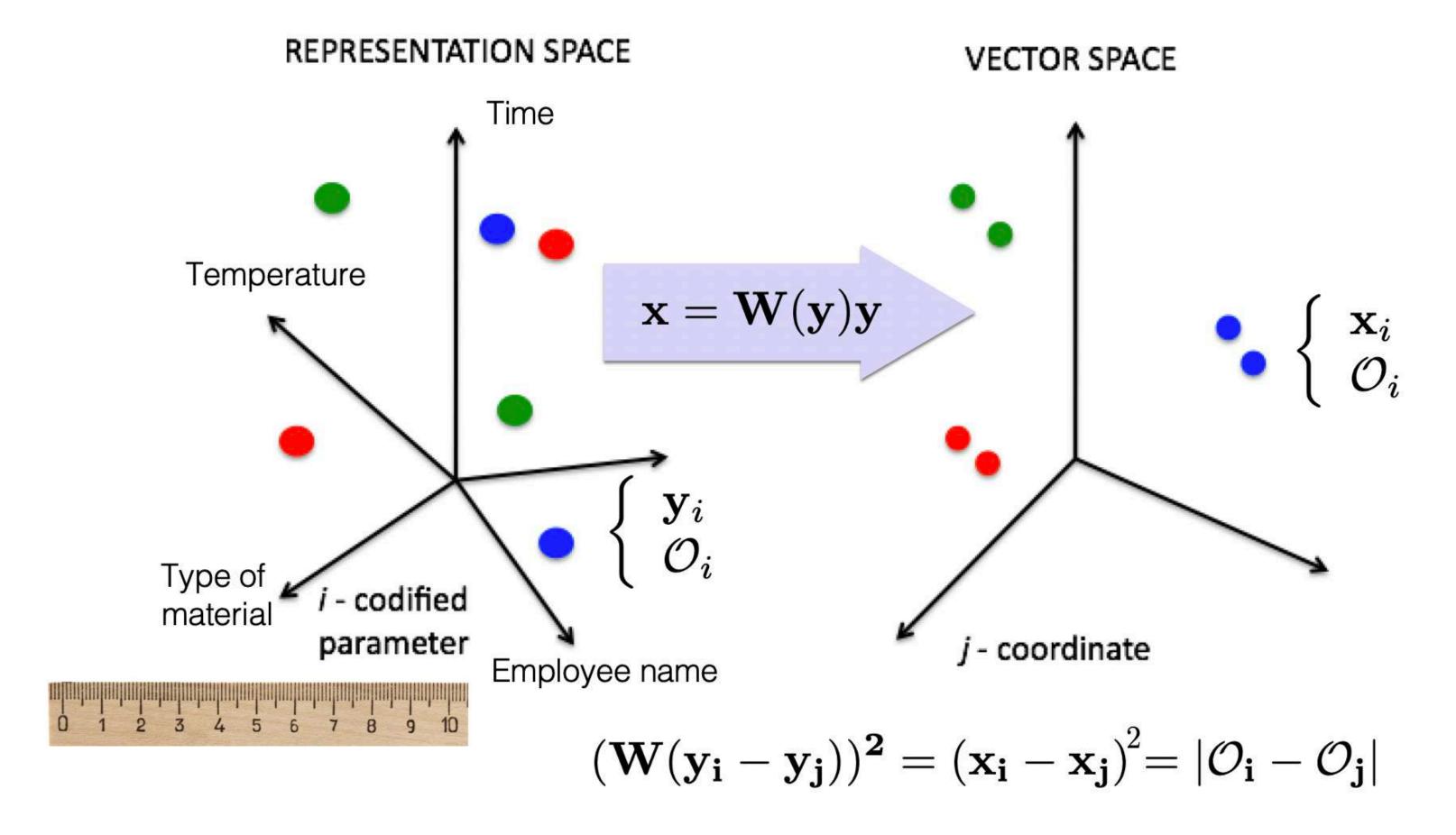




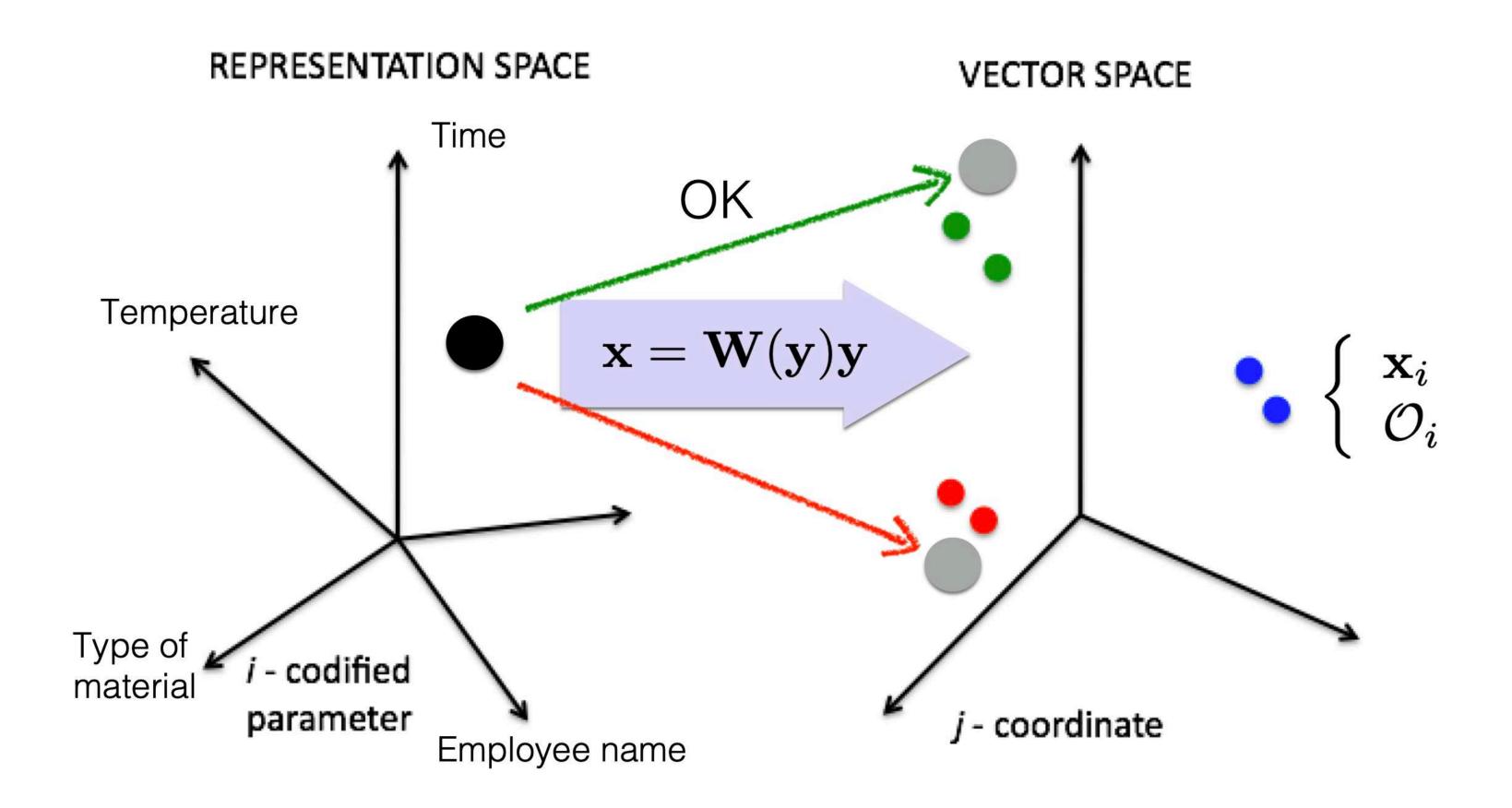


- Learning metrics
- Data transformation
- Extracting data features

Learning metrics: e.g. Code2Vect

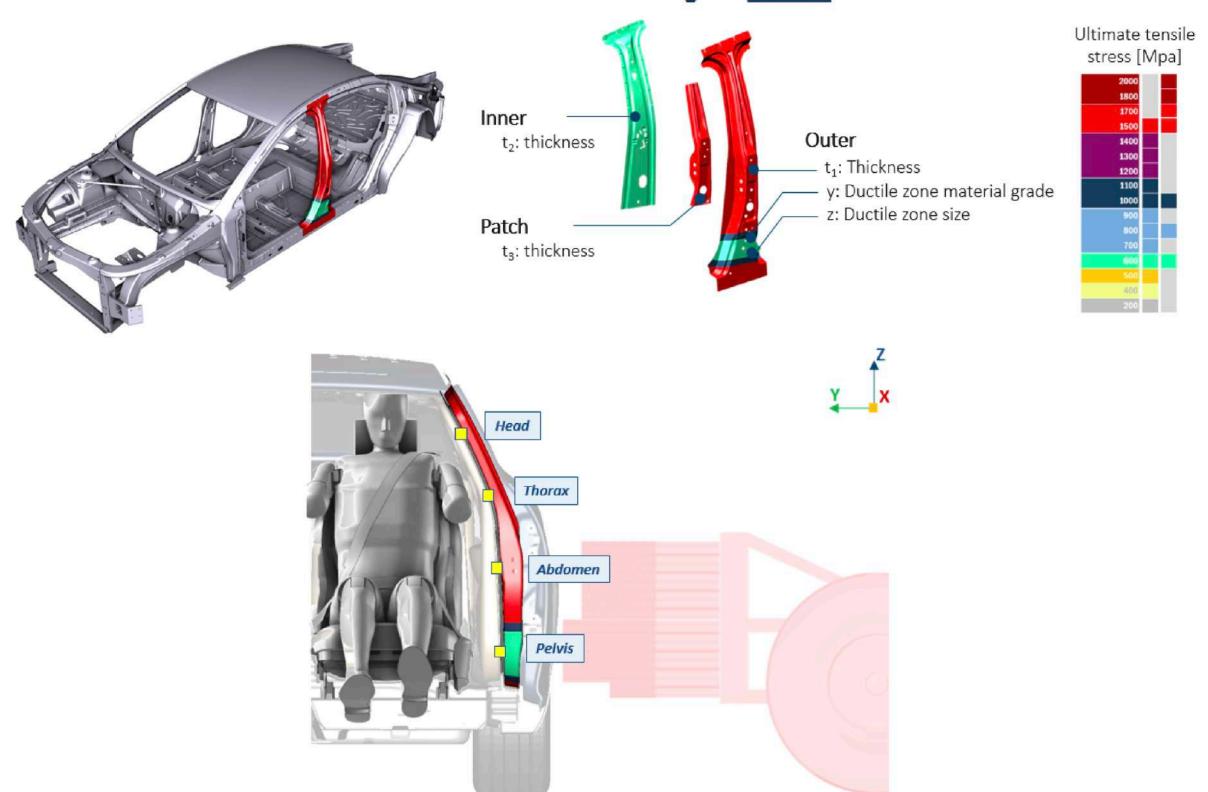


Code2Vect



Code2Vect

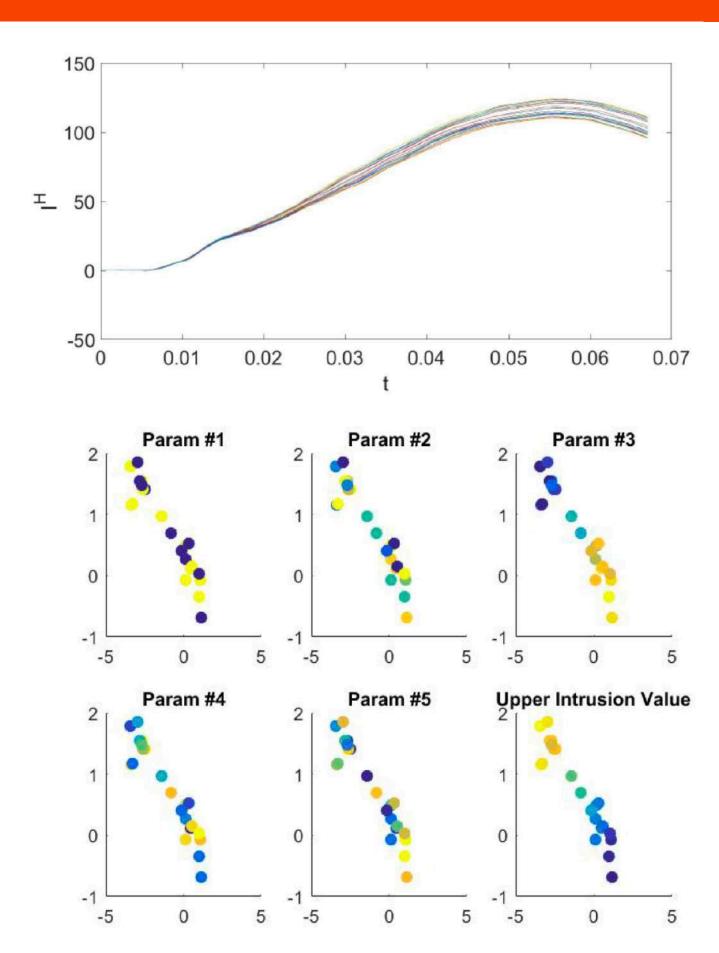




Code2Vect

DoE & High Fidelity simulations

| Simulation | Z | t_2 | $\mathbf{t_1}$ | $\mathbf{t_3}$ | y |
|------------|----|-------|----------------|----------------|-----|
| 1 | 90 | 1.3 | 1.5 | 1.5 | 471 |
| 2 | 90 | 1.15 | 1.7 | 1.5 | 591 |
| 3 | 60 | 1.25 | 1.7 | 1.1 | 549 |
| 4 | 60 | 0.9 | 1.65 | 1.05 | 527 |
| 5 | 60 | 1.25 | 1.55 | 1.15 | 386 |
| 6 | 60 | 1.3 | 1.15 | 1.55 | 356 |
| 7 | 90 | 1.2 | 1.15 | 1.4 | 570 |
| 8 | 90 | 1.25 | 1.7 | 1 | 364 |
| 9 | 90 | 1.1 | 1.75 | 1.1 | 591 |
| 10 | 90 | 1 | 1.1 | 1.05 | 420 |
| 11 | 90 | 0.95 | 1.15 | 1.55 | 527 |
| 12 | 90 | 0.9 | 1.65 | 1.55 | 489 |
| 13 | 90 | 1.2 | 1.2 | 1.6 | 360 |
| 14 | 60 | 0.95 | 1.6 | 1.1 | 351 |
| 15 | 60 | 1.1 | 1.35 | 1.5 | 557 |
| 16 | 90 | 1.1 | 1.4 | 1.25 | 347 |
| 17 | 90 | 1.3 | 1.1 | 1.1 | 493 |
| 18 | 90 | 1.1 | 1.65 | 1.55 | 390 |
| 19 | 60 | 1.3 | 1.1 | 1.25 | 471 |
| 20 | 60 | 0.9 | 1.15 | 1.2 | 536 |
| 21 | 60 | 1 | 1.2 | 1.35 | 394 |
| 22 | 60 | 1.3 | 1.6 | 1.6 | 531 |



Beyond Euclidian Metrics

Apparently three trees, apparently!



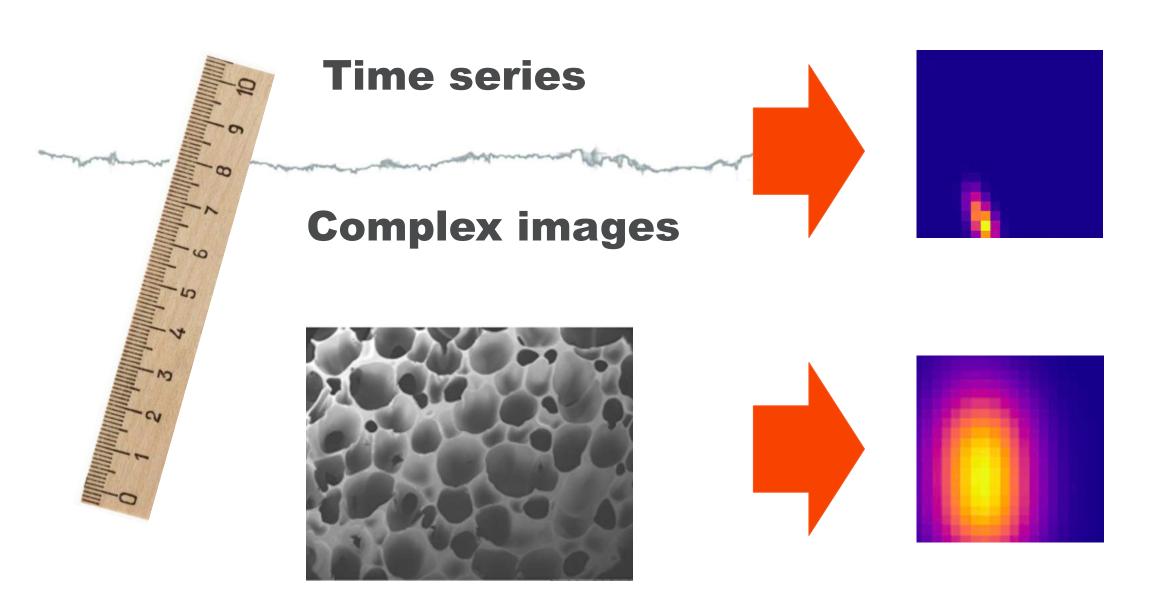




In what sense they are close?
What kind of resemblance?
How many parameters define them?
What is the adequate metric for comparing them?

Data transformation, e.g. TDA

Topological Data Analysis - TDA



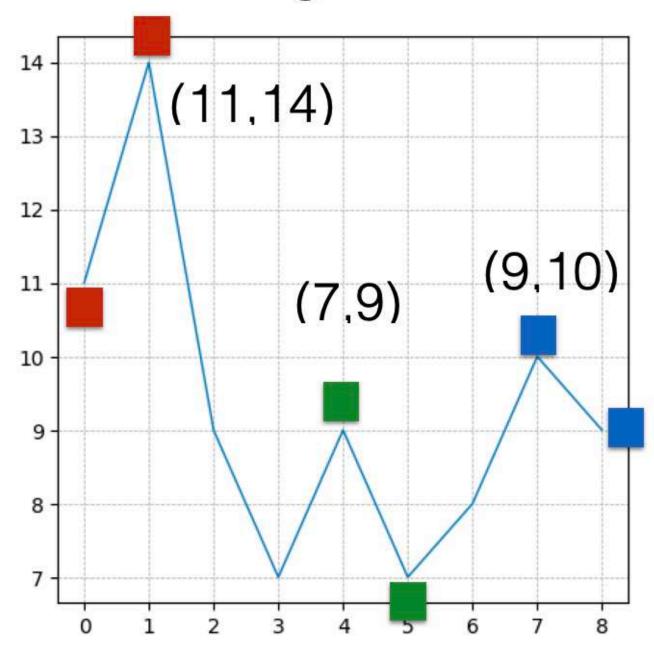
A sort of goal oriented

QR code / Passeport

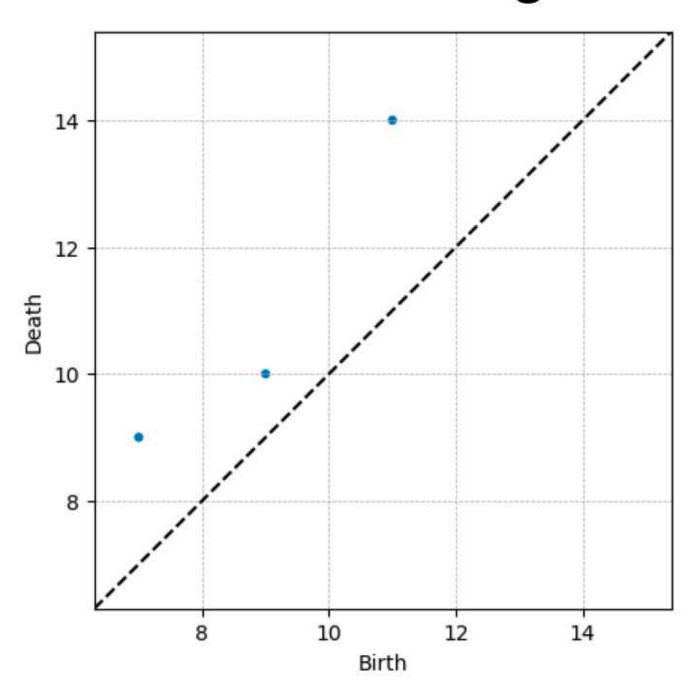


BUT in a vector space

Pairing min-max

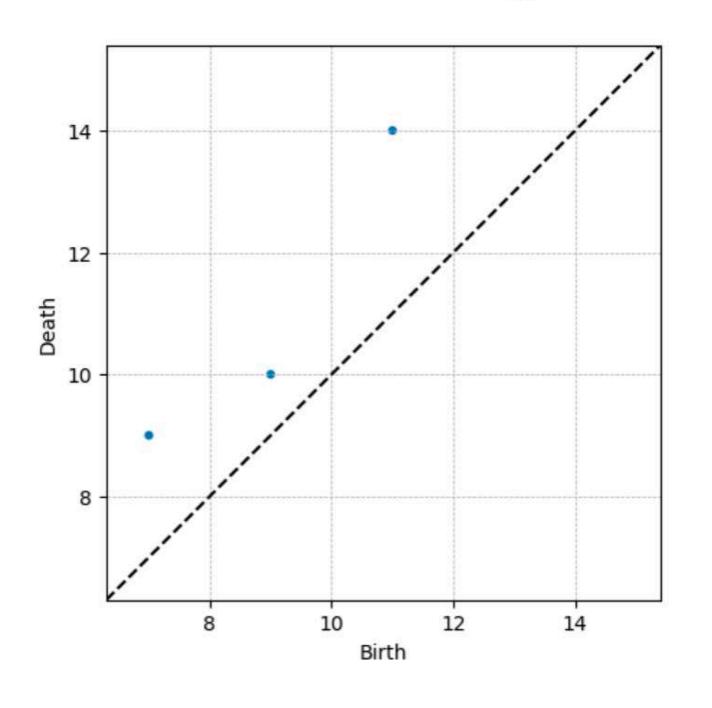


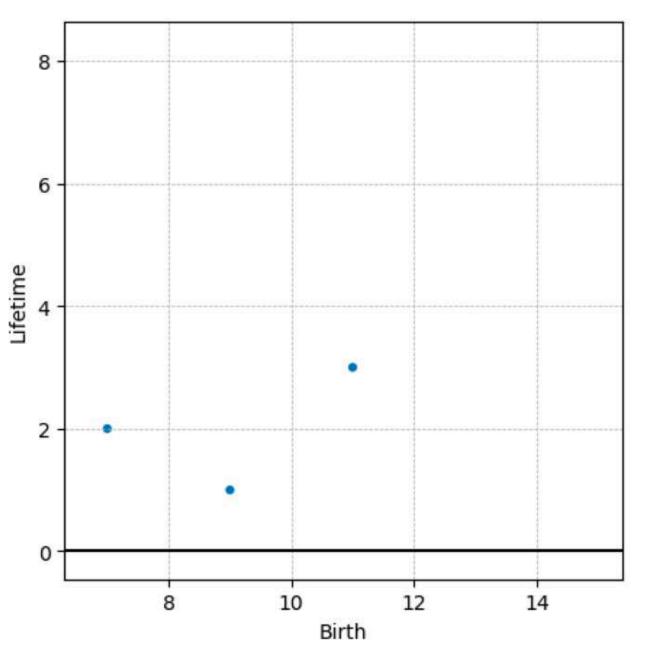
Persistence diagram



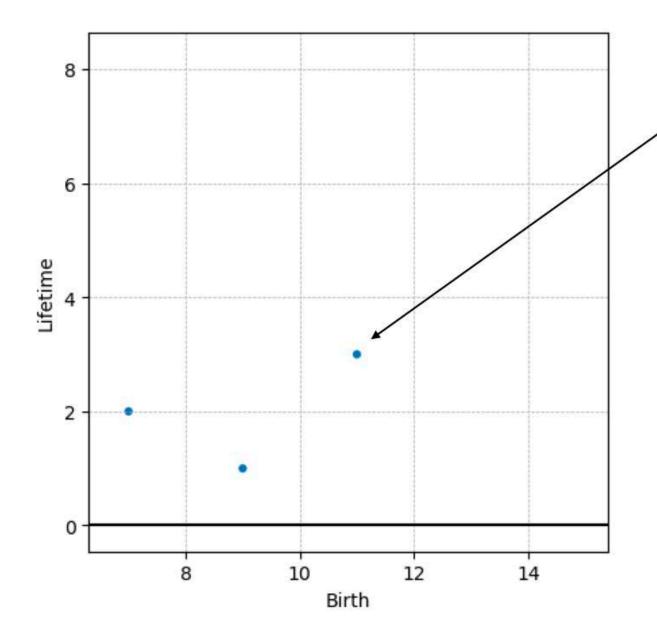
Persistence diagram

Lifetime diagram



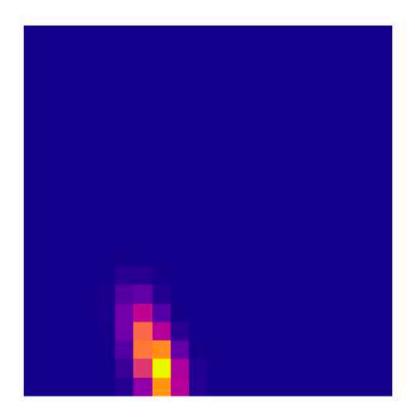


Lifetime diagram

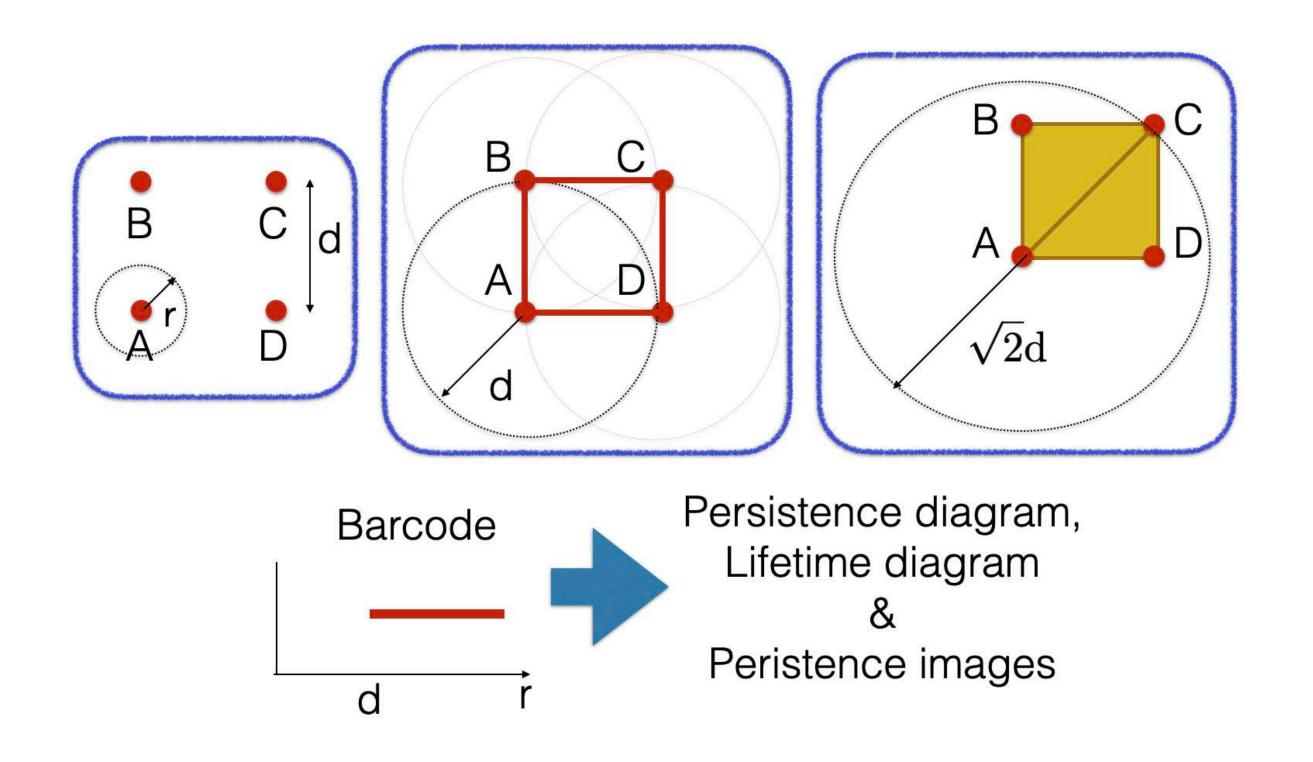


$$\rho_S(u,v) = \sum_{(x,y)\in\mathcal{T}(S)} w(x,y) g_{(x,y)}(u,v),$$

$$\mathcal{P}I_{P_i}(S) = \iint_{P_i} \rho_S(u, v) du dv.$$

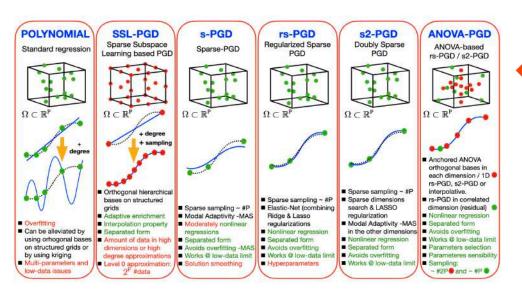


Persistence image



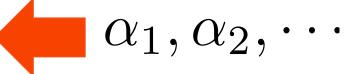


Regressions PGD-based



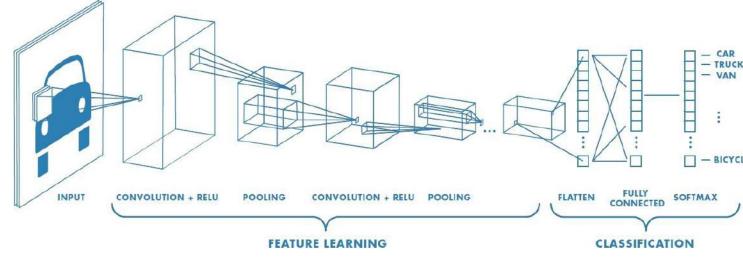


Reduction POD





(convolutional) Neural Network





PERFORMANCES

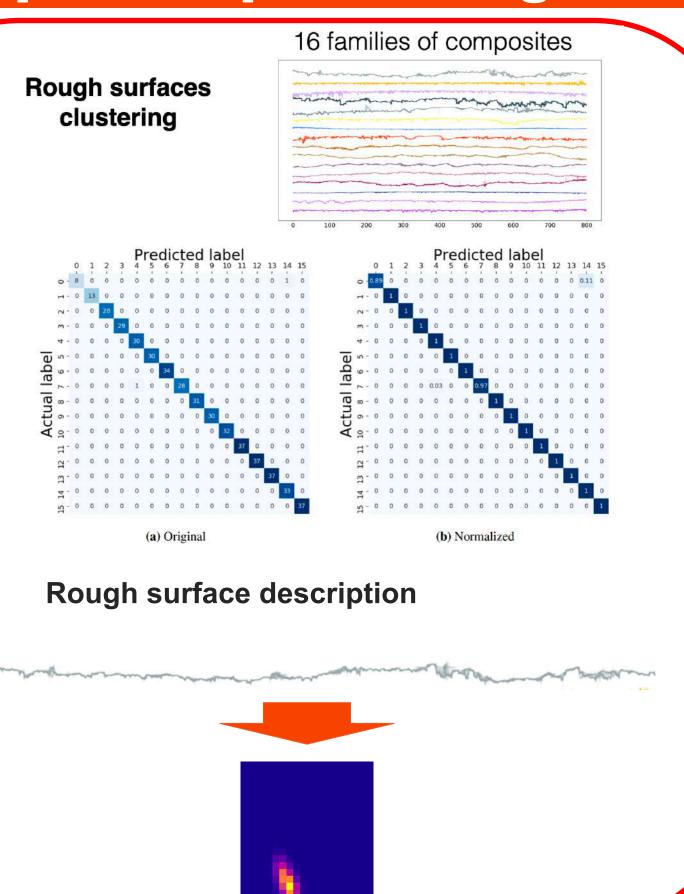




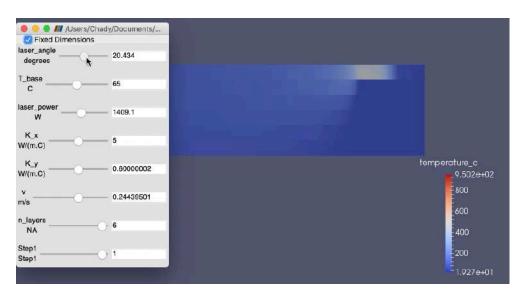
1. Introduction

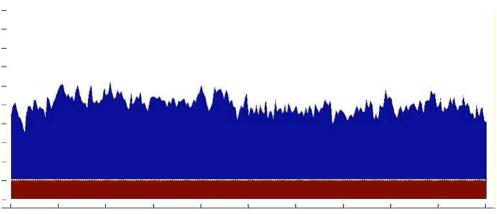
2. Applications

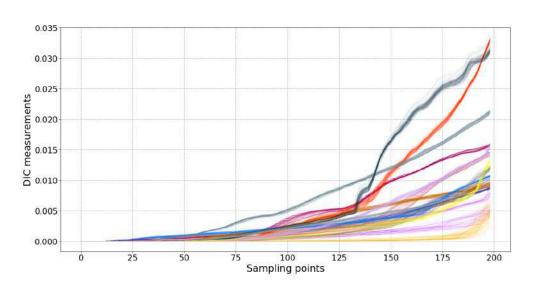
Composites processing

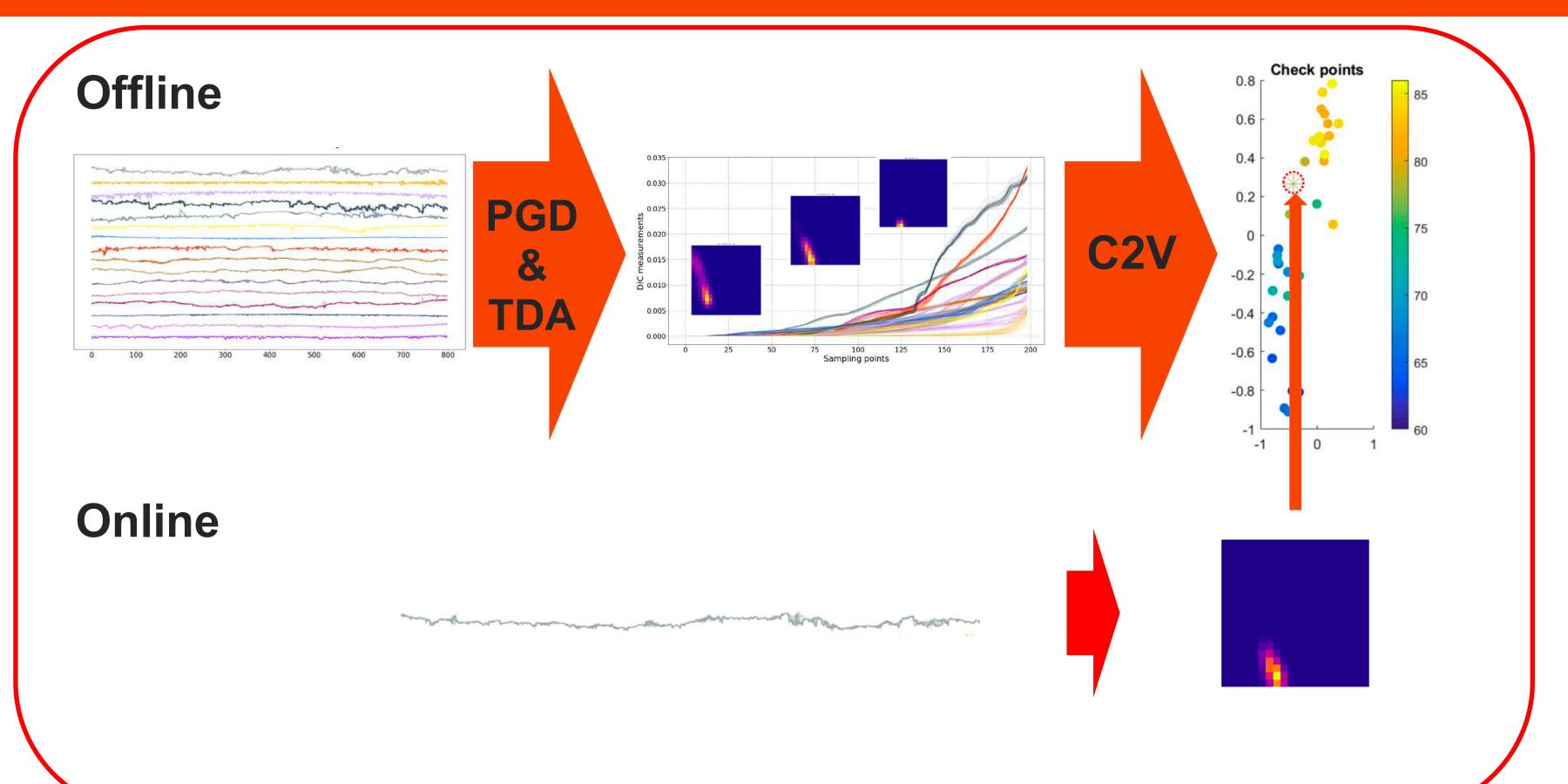


Rough surfaces consolidation analysis

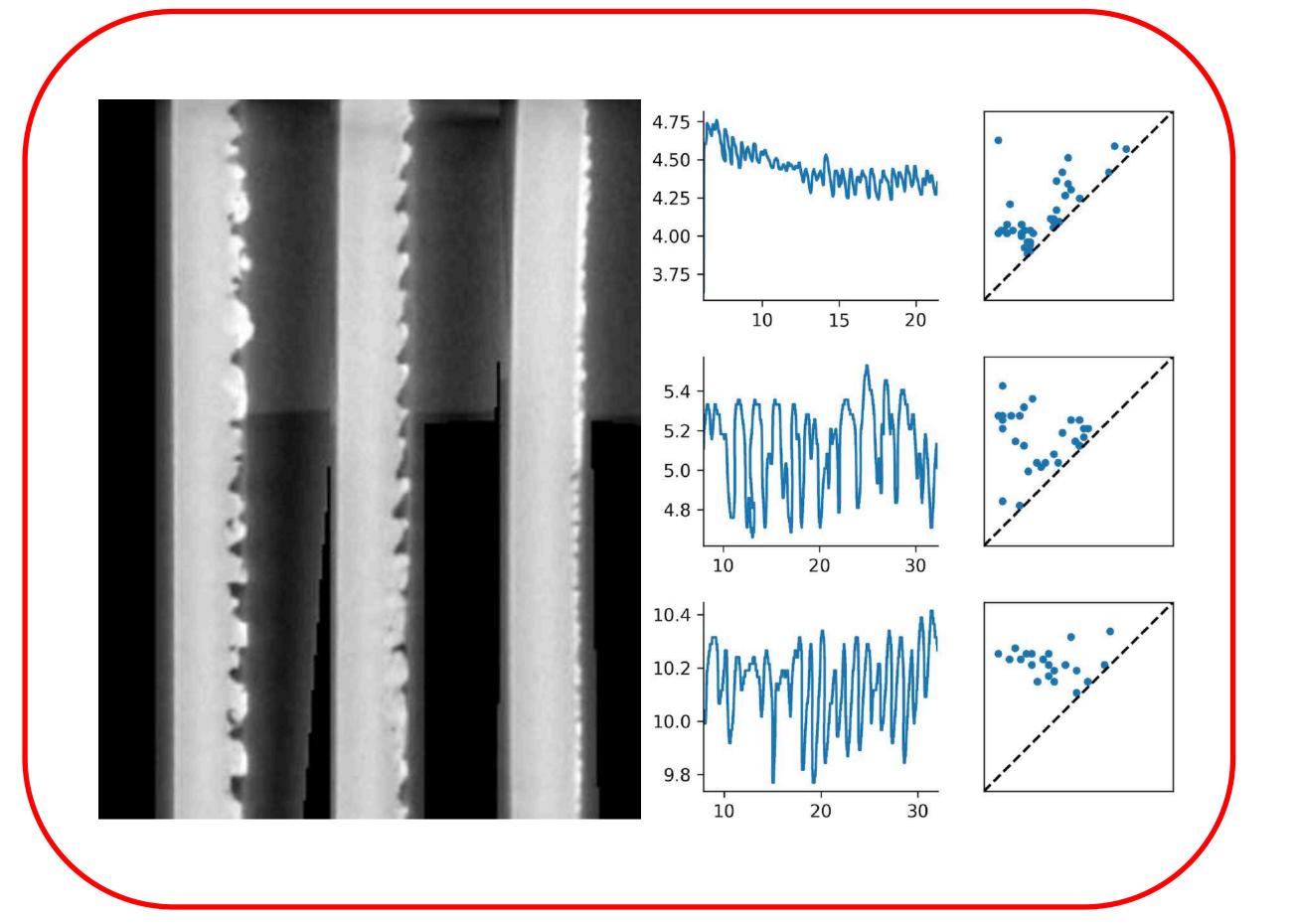




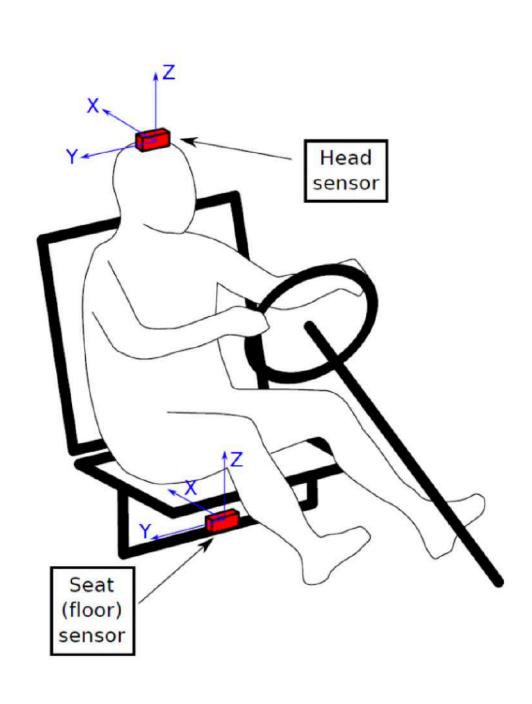




Bonding @ laser structured surfaces

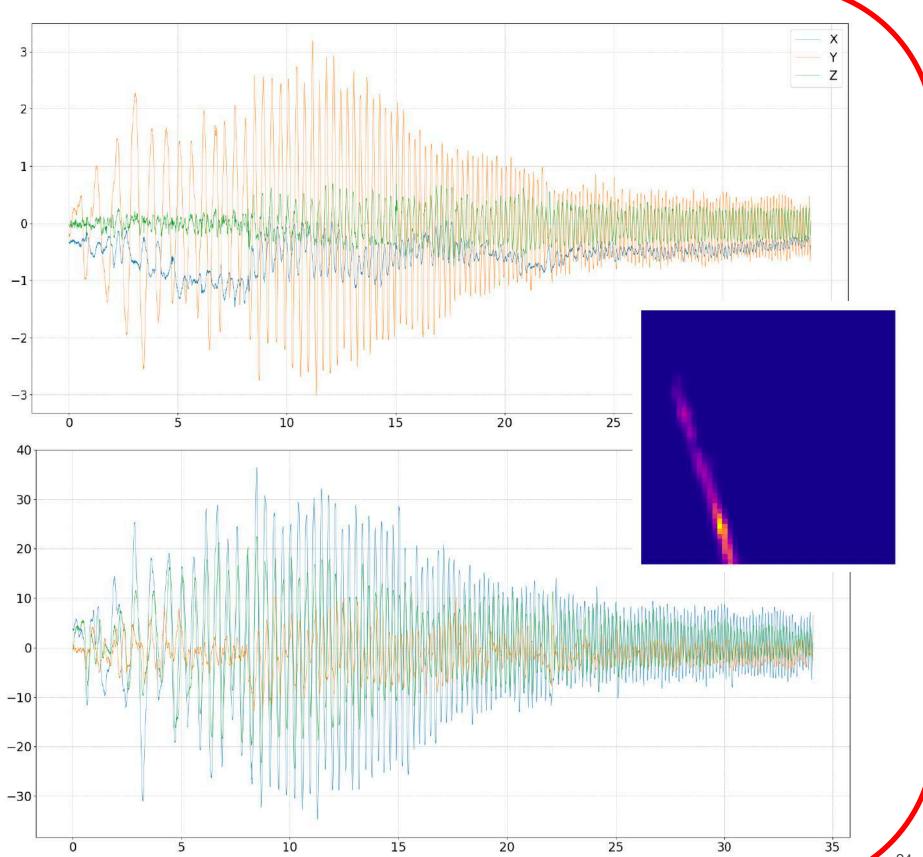


ADAS

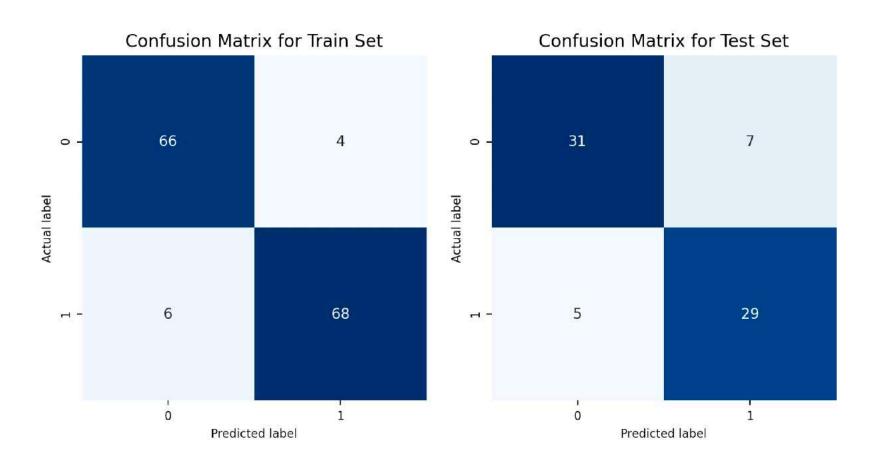




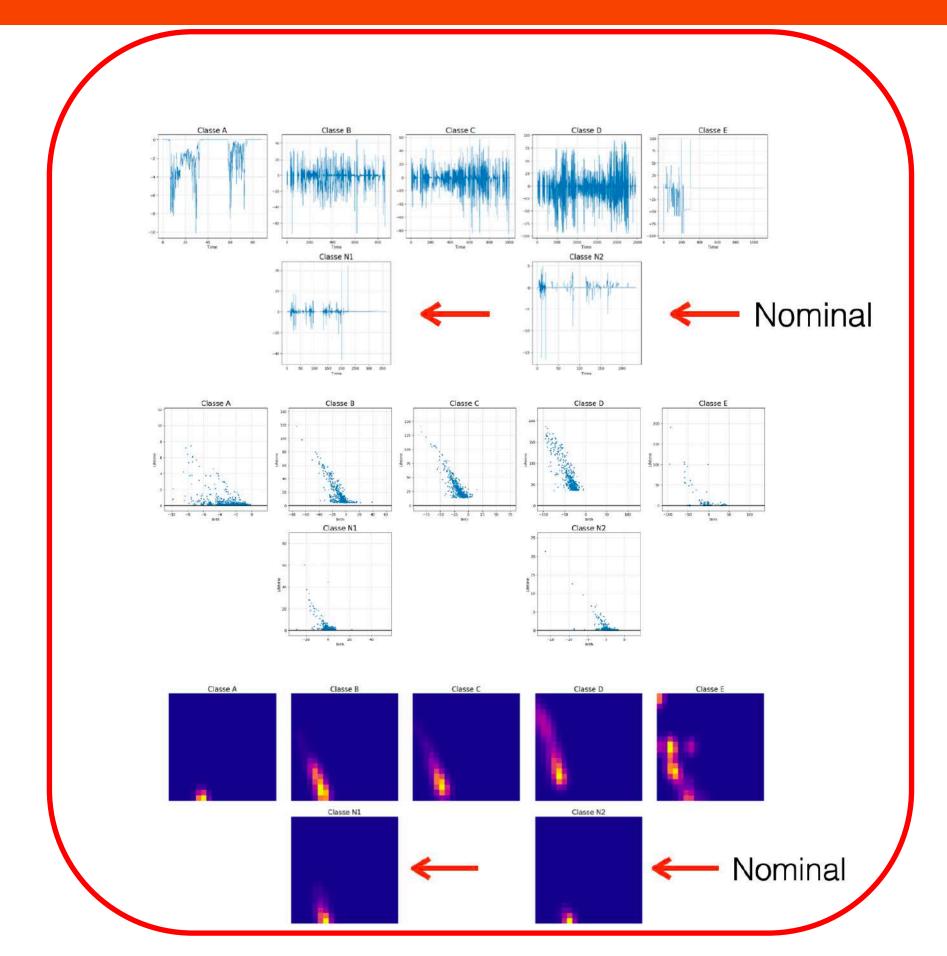




Class Label SAVRelaxedPassager SAVTensePassager SAVRelaxedDriver SAVTenseDriver 5 RigidRelaxedPassager RigidTensePassager RigidRelaxedDriver RigidTenseDriver

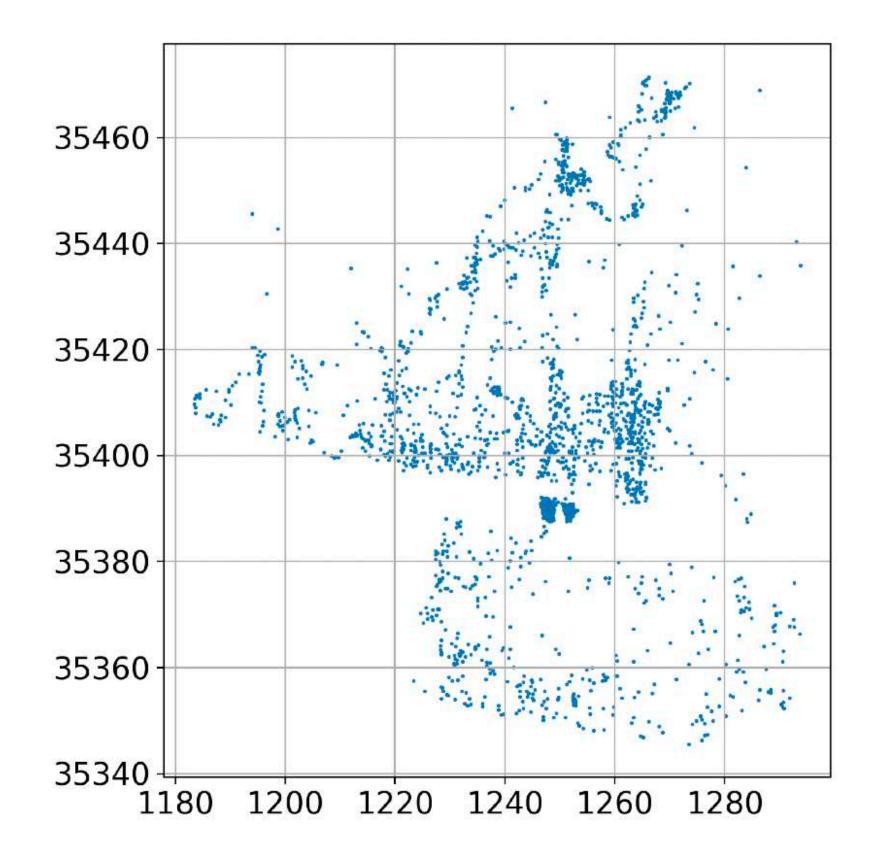


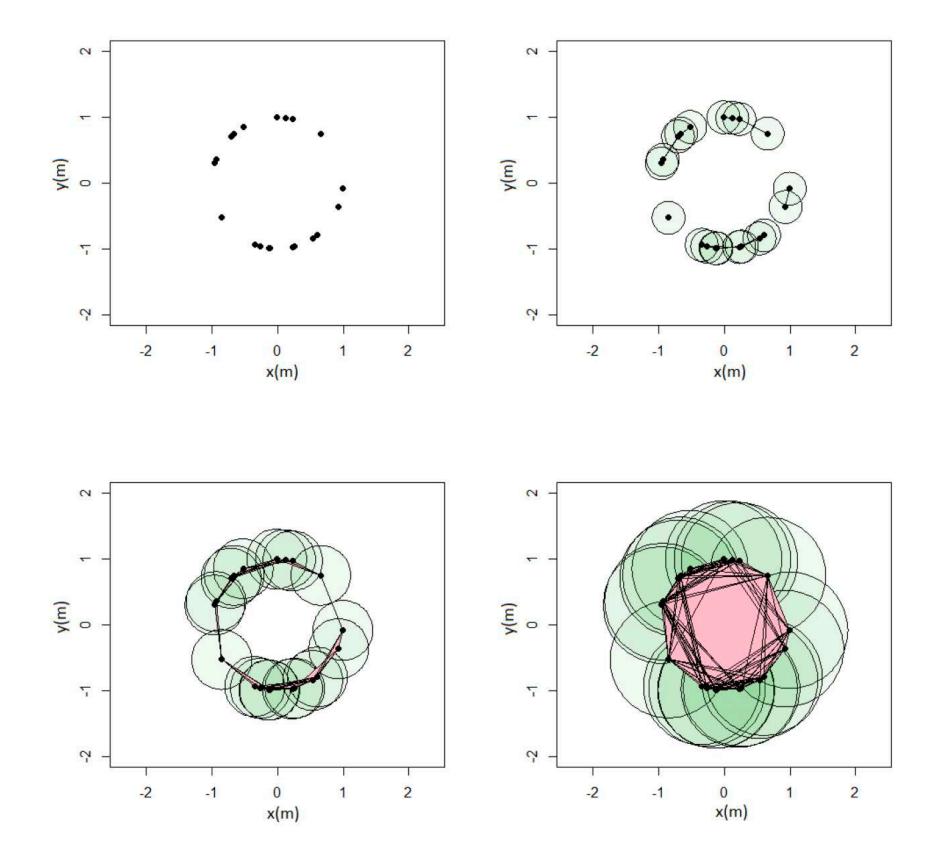
FAULTS

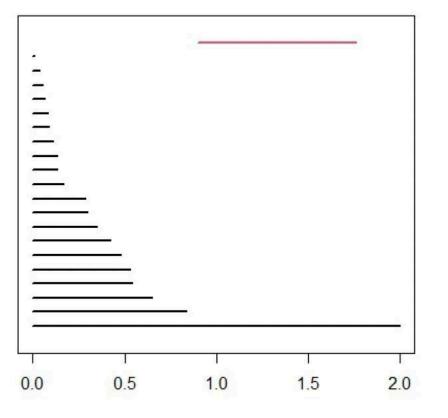


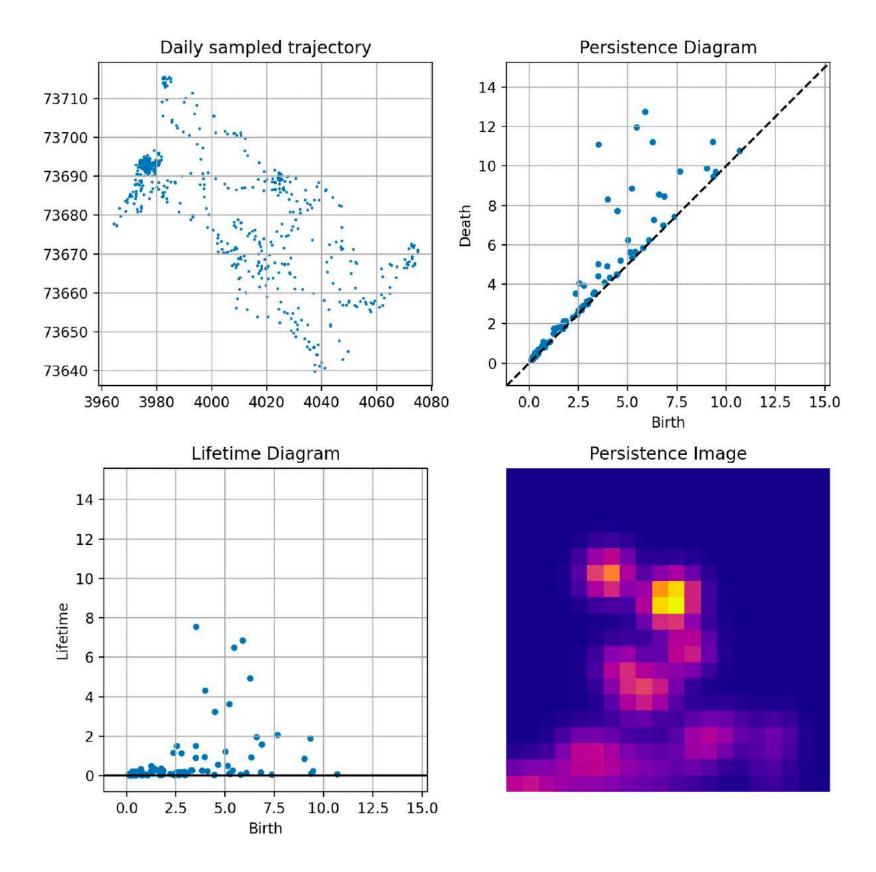
Robots

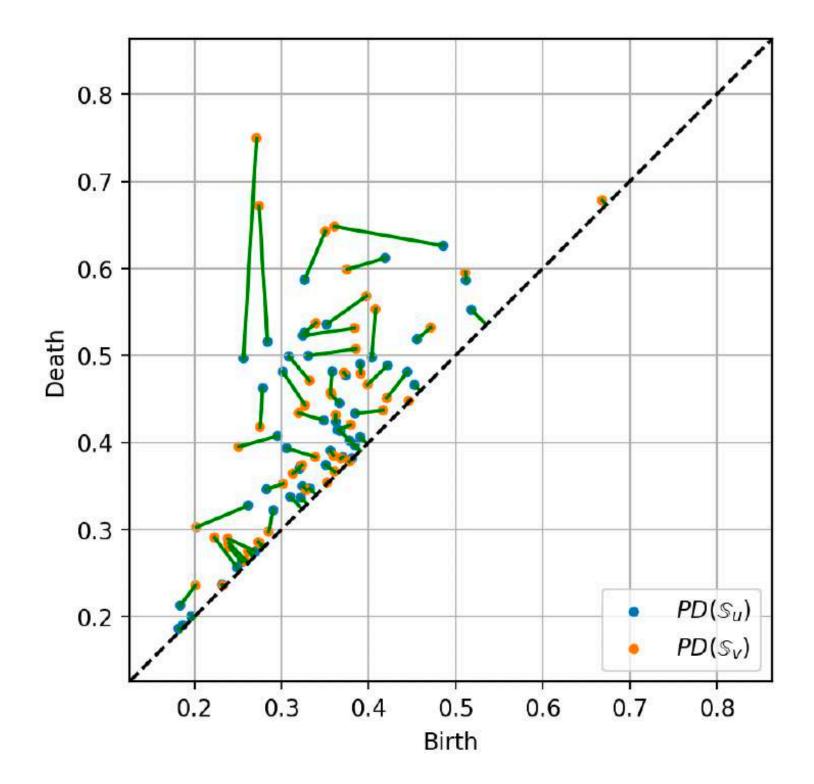


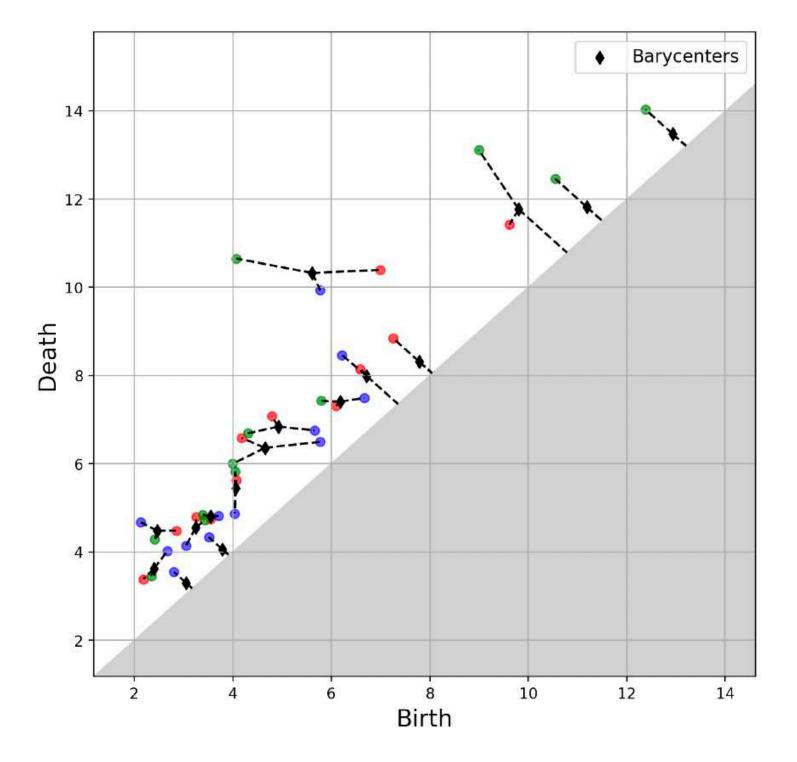


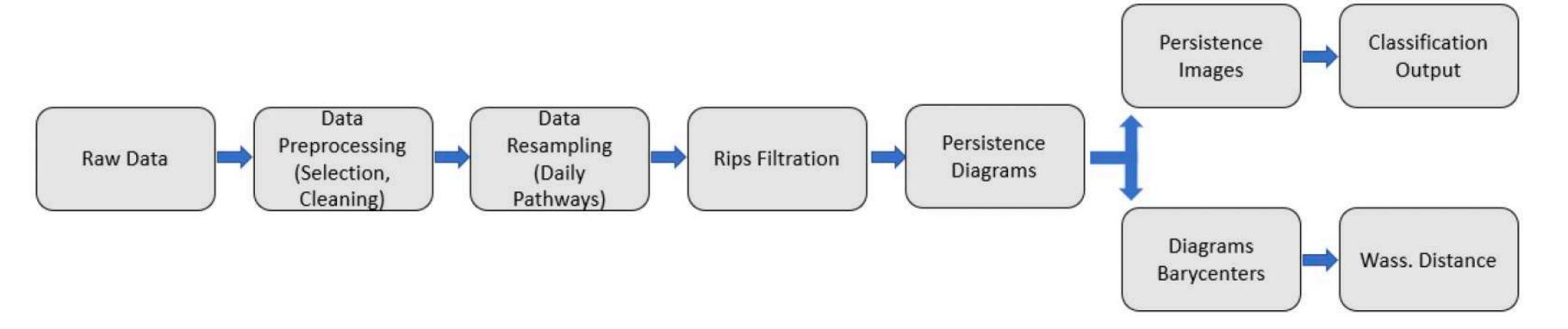


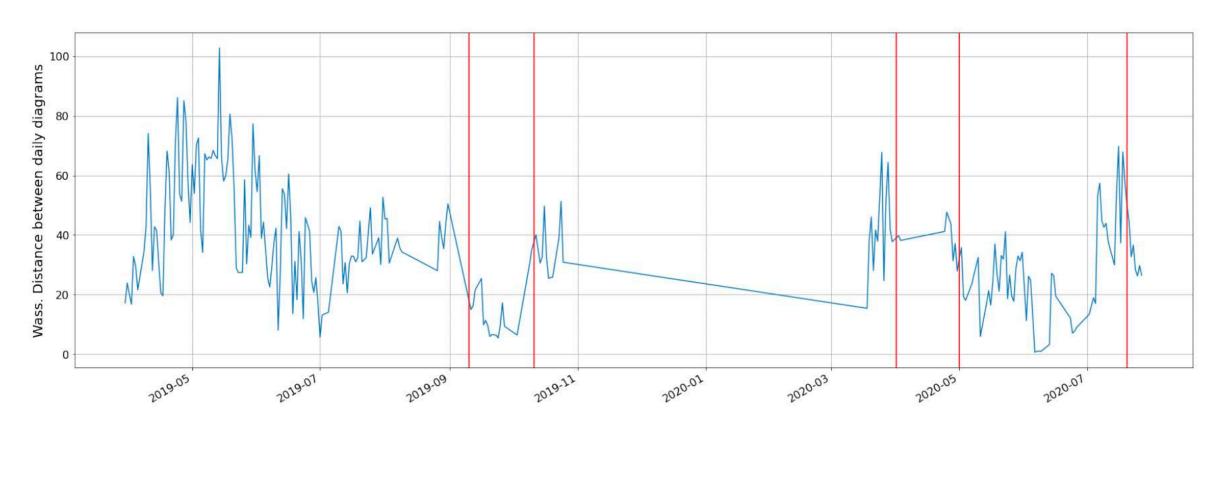


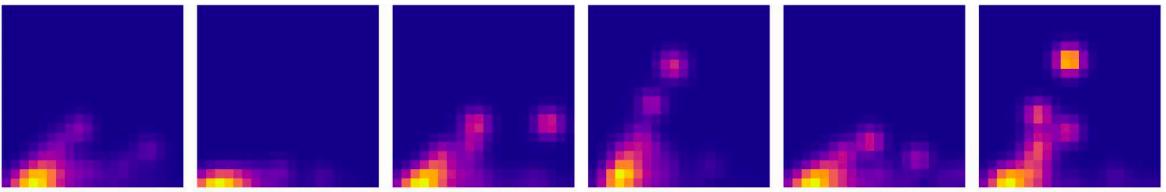




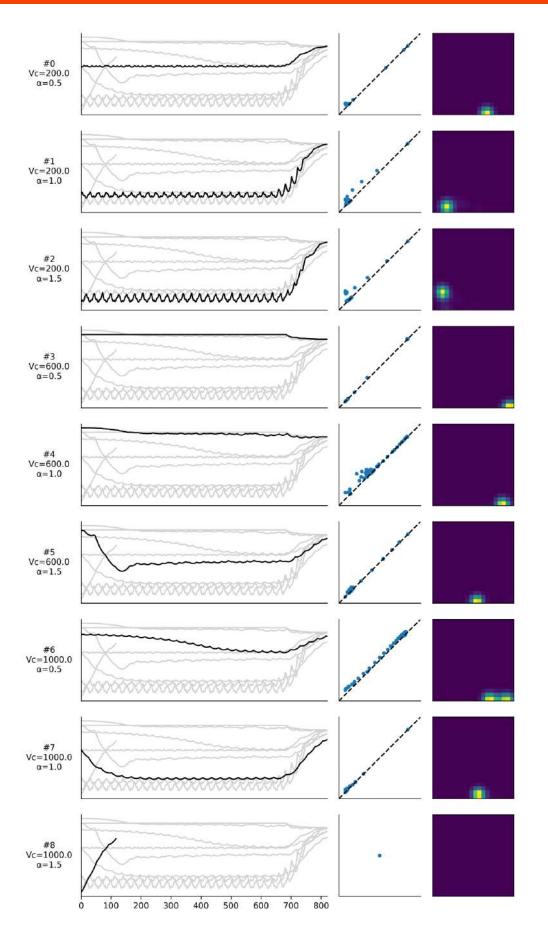








Surface quality in machining



POD on the images

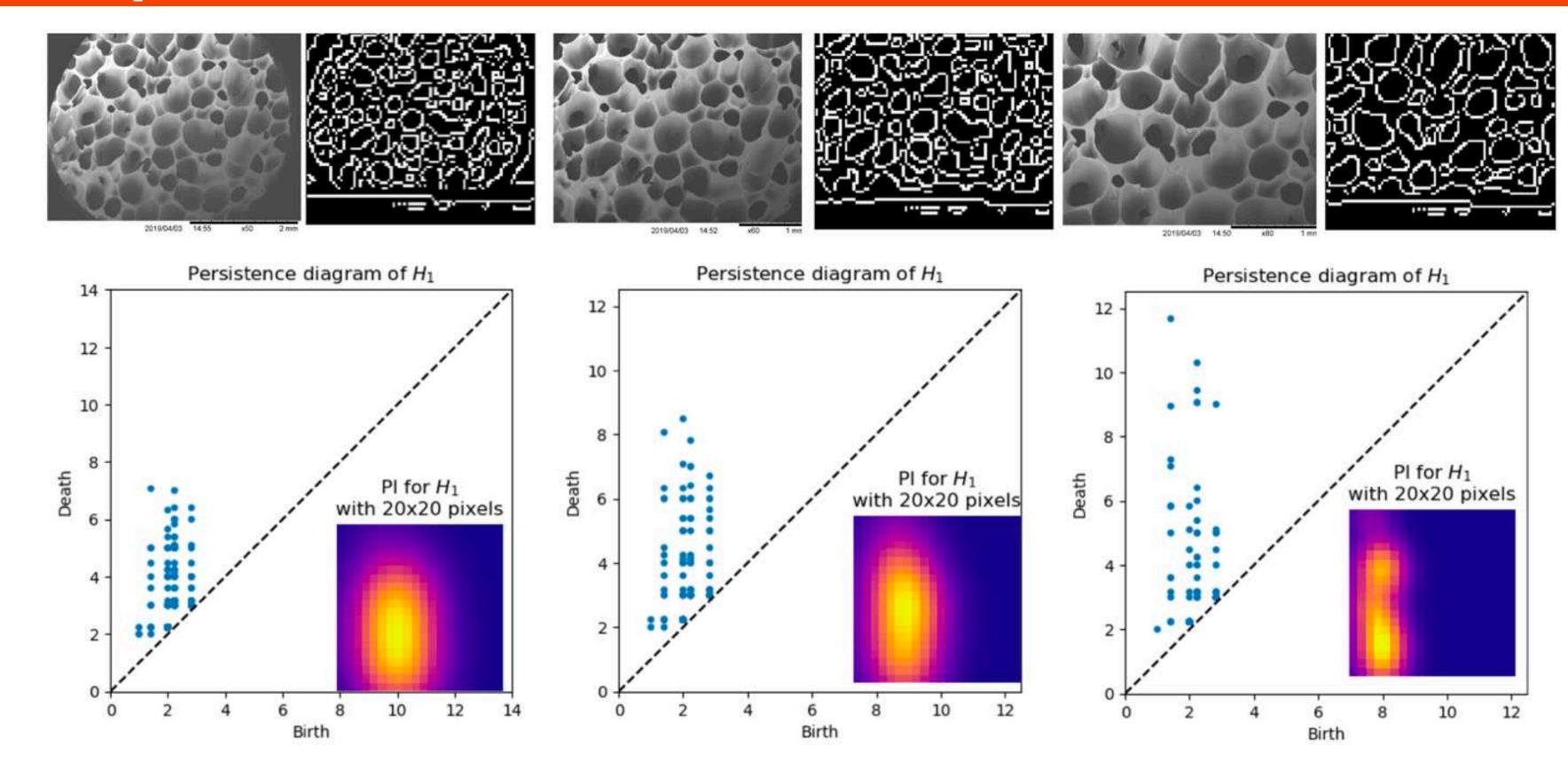
$$PI = \sum_{i=1}^{R} \alpha_i(\mu_1, \mu_2) M_i$$

 (μ_1, μ_2) process parameters M_i POD modes

$$(\mu_1^*, \mu_2^*) \to \alpha_i^*, i = 1, ..., R \to PI^*$$

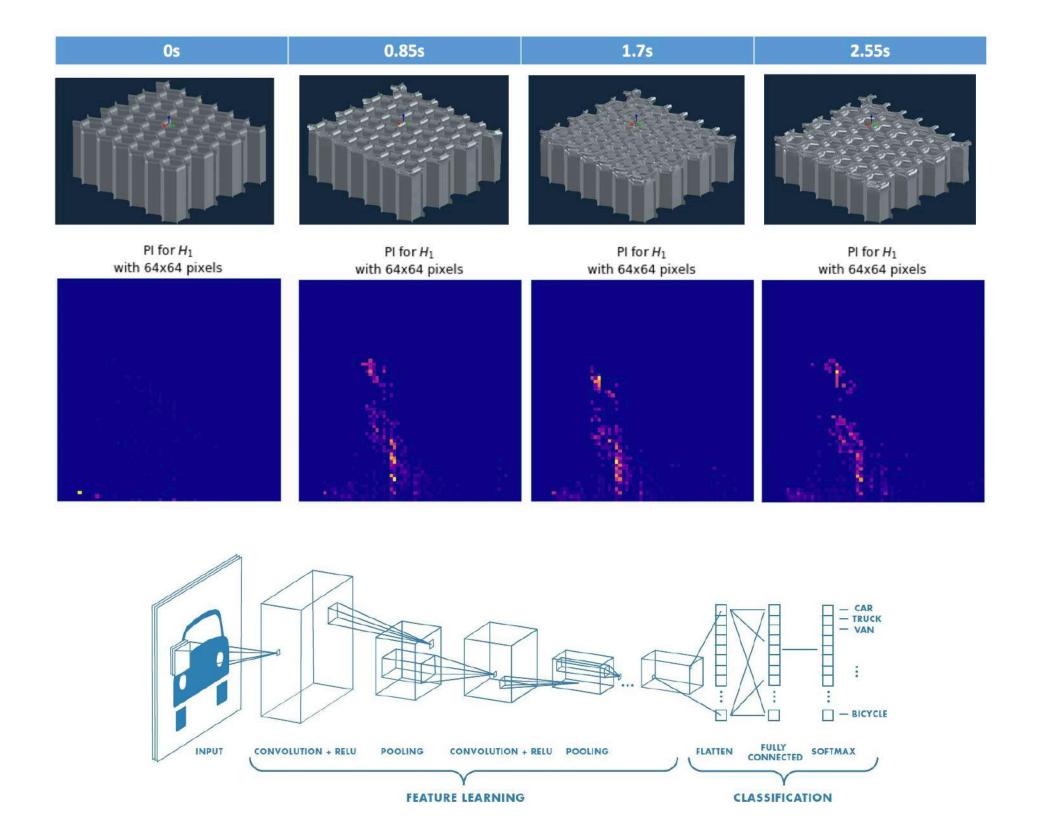
Look for the closest PI_j to PI^* that represents the expected surface

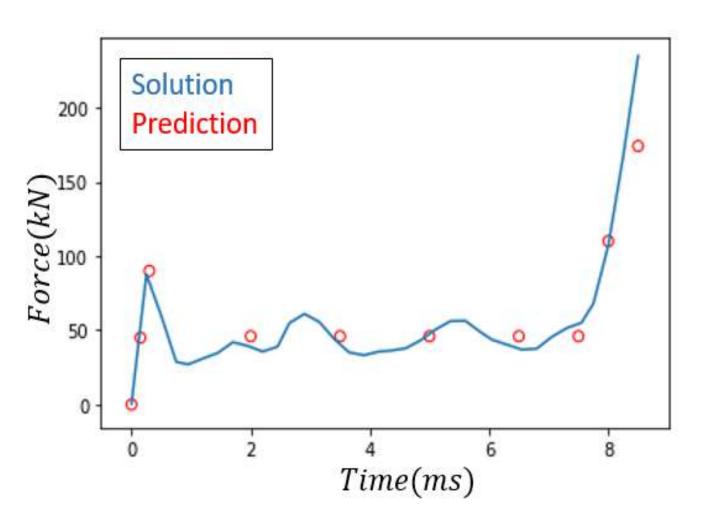
Complex microstructures



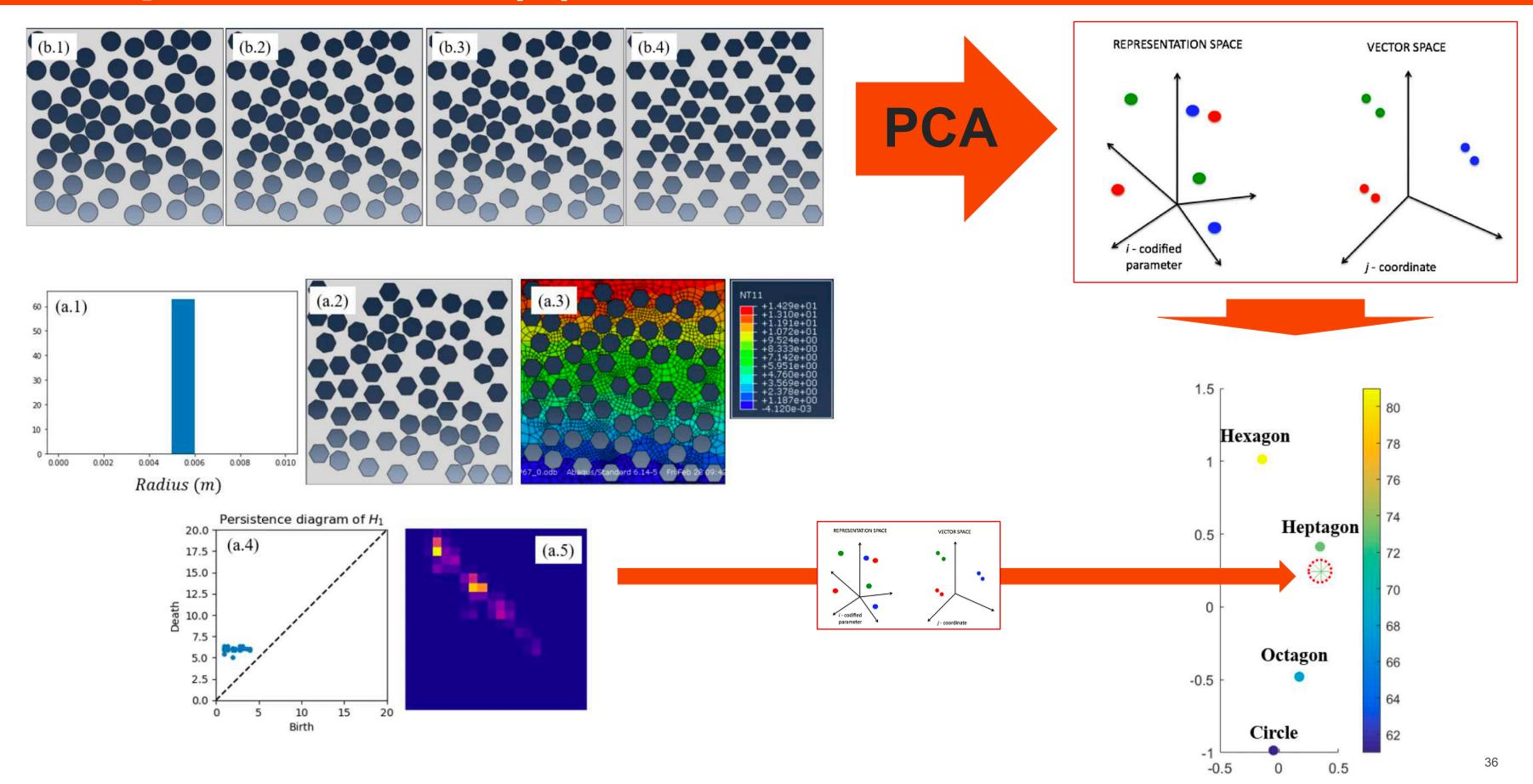
Complex behaviors

Data-driven modelling of honeycomb barriers

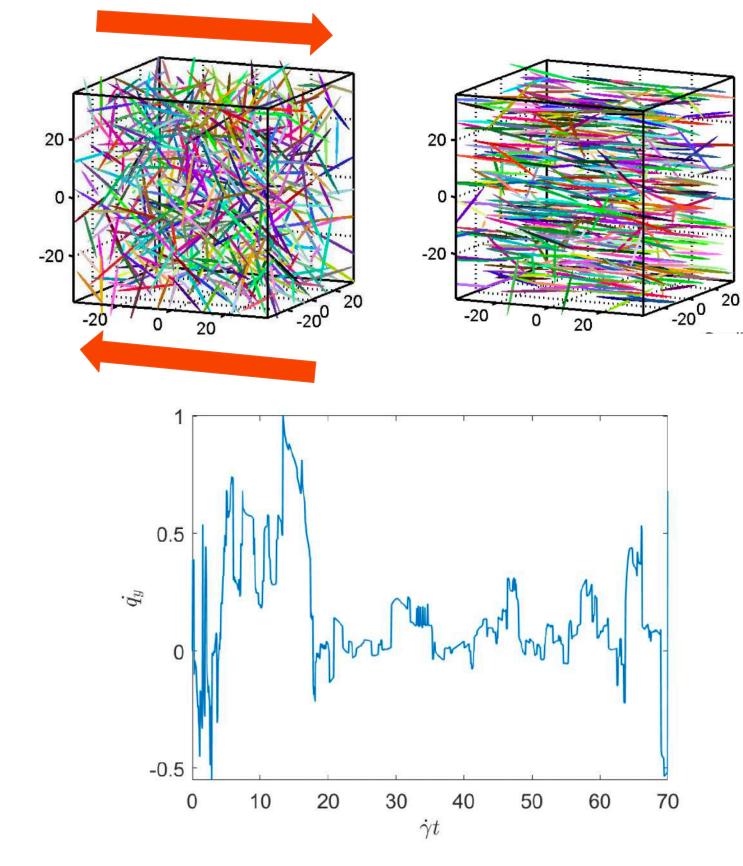




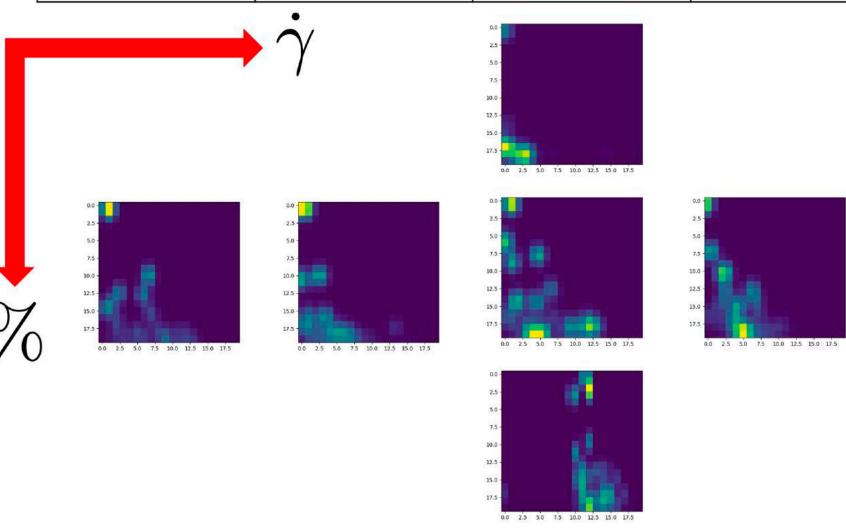
Complex behaviors (II)



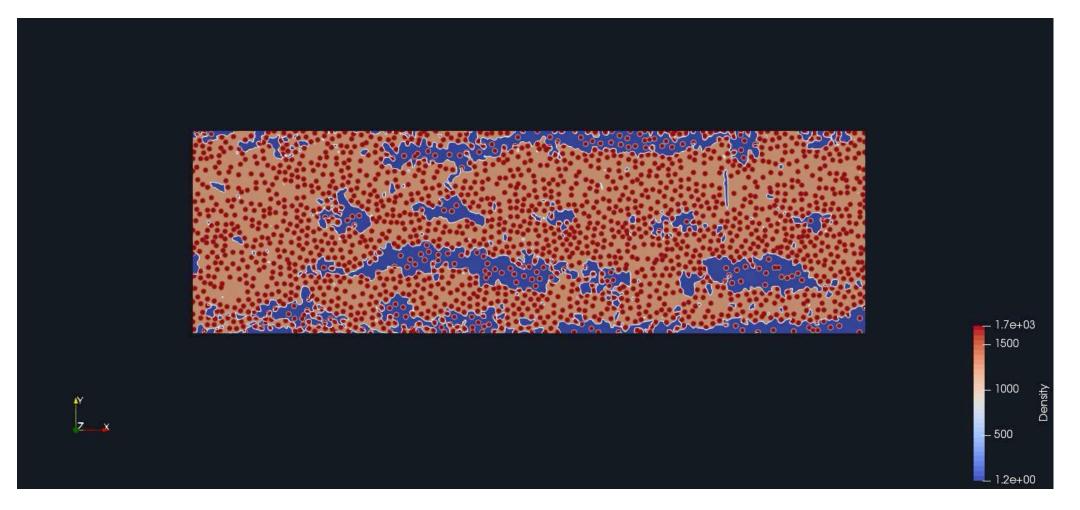
Rheology

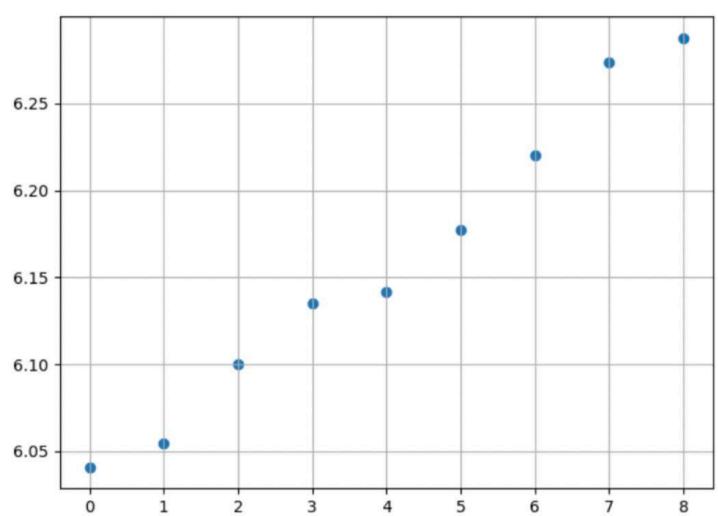


| | | Case 1: 5s ⁻¹ , 14% | |
|---------------------------------------|---------------------------------------|---------------------------------------|--------------------------------|
| Case 4: 1s ⁻¹ , 18% | Case 5: 3s ⁻¹ , 18% | Case 2: 5s ⁻¹ , 18% | Case 6: 7s ⁻¹ , 18% |
| | | Case 3: 5s ⁻¹ , 22% | |



Rheology (II)

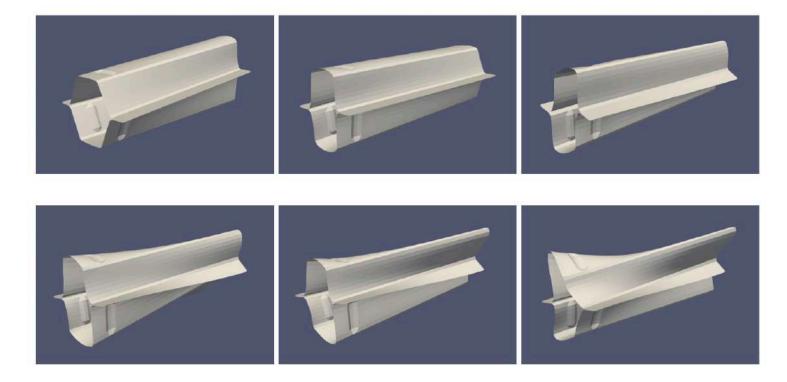




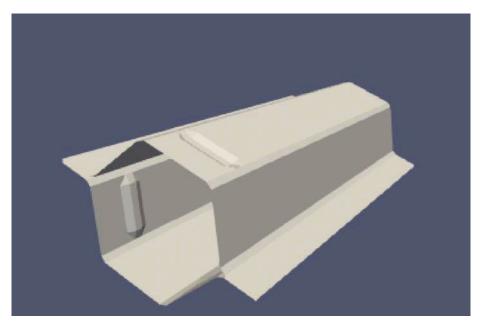
Parametric vibration modes clustering for advanced NVH

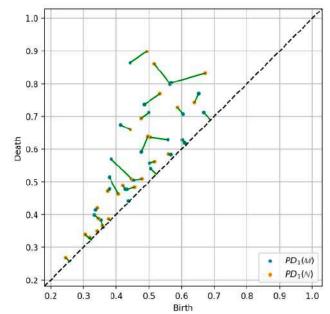
17 structure configurations: different thickness.

First 6th modes for a given structure thickness:



Use of alpha-filtration and Wasserstein distance (with respect to the undeformed geometry) for modes classification





| Case | 1st surf. | 2nd surf. | 3rd surf. | 4th surf. | 5th surf. | 6th surf. |
|------|-----------|-----------|-----------|-----------|-----------|-----------|
| 01 | 1 | 2 | 3 | 4 | 5 | 6 |
| 02 | 1 | 2 | 3 | 4 | 5 | 6 |
| 03 | 1 | 2 | 3 | 4 | 5 | 6 |
| 04 | 1 | 2 | 3 | 5 | 4 | 6 |
| 05 | 1 | 2 | 3 | 4 | 5 | 6 |
| 06 | 1 | 2 | 3 | 5 | 4 | 6 |
| 07 | 1 | 2 | 3 | 4 | 5 | 6 |
| 08 | 1 | 2 | 3 | 4 | 5 | 6 |
| 09 | 1 | 2 | 4 | 3 | 6 | 5 |
| 10 | 2 | 1 | 3 | 4 | 6 | 5 |
| 11 | 1 | 2 | 3 | 4 | 5 | 6 |
| 12 | 1 | 2 | 3 | 4 | 5 | 6 |
| 13 | 1 | 2 | 3 | 5 | 4 | 6 |
| 14 | 1 | 2 | 4 | 3 | 6 | 5 |
| 15 | 1 | 2 | 4 | 3 | 5 | 6 |
| 16 | 1 | 2 | 3 | 4 | 5 | 6 |
| 17 | 1 | 2 | 3 | 4 | 6 | 5 |

REFERENCES

- Nonlinear regression operating on microstructures described from Topological Data Analysis for the real-time prediction of effective properties. M. Yun, C. Argerich, E. Cueto, J.L. Duval, F. Chinesta. Materials, 13/10, 2335, 2020.
- Tape Surfaces Characterization with Persistence Images. T. Frahi, M. Yun, C. Argerich, A. Falco, F. Chinesta. AIMS Materials Science, 7/4, 364-380, 2020
- Empowering Advanced Driver-Assistance Systems from Topological Data Analysis. T. Frahi, F.
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- Monitoring weeder robots and anticipating their functioning by using advanced topological data analysis. T. Frahi, A. Sancarlos, M. Galle, X. Beaulieu, A. Chambard, A. Falco, E. Cueto, F. Chinesta. Frontiers in Artificial Intelligence.
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 Mezher, J. Arayro, N. Hascoet, F. Chinesta. Entropy.