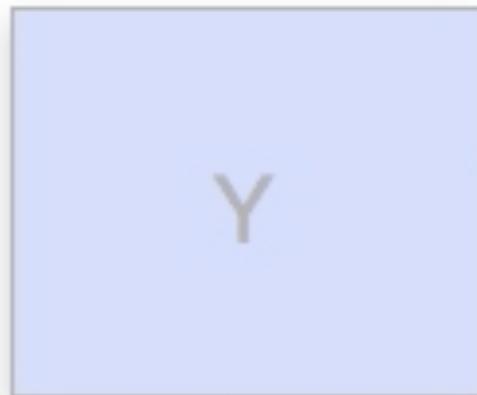


Gromov-Hausdorff stable signatures for shapes using persistence

joint with F. Chazal, D. Cohen-Steiner, L. Guibas and S. Oudot

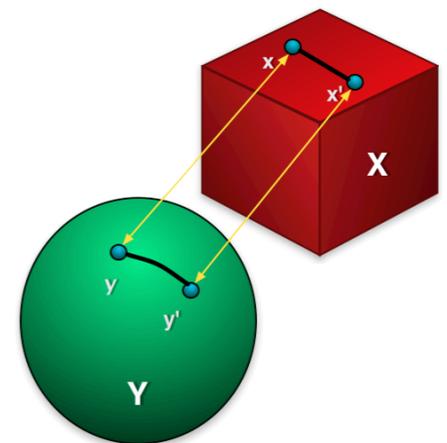


Facundo Mémoli
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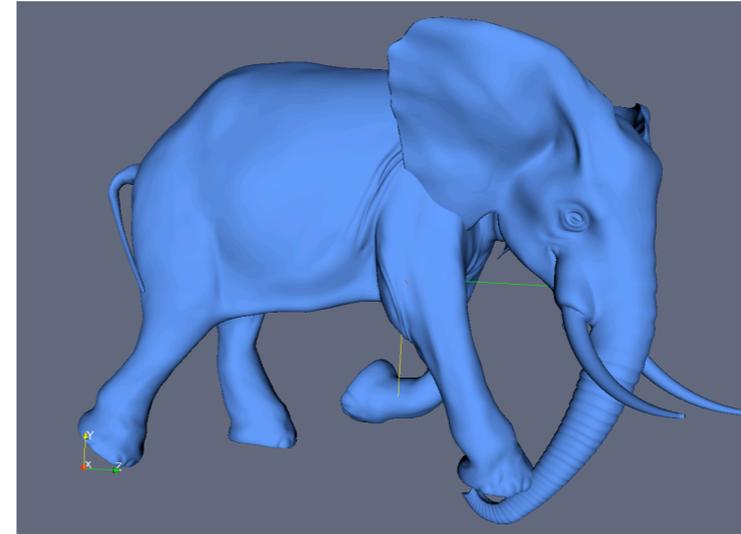
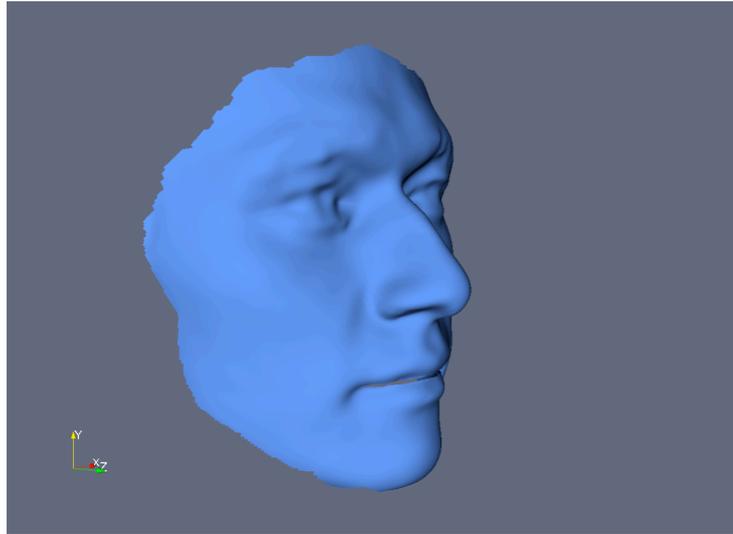
Goal

- Shape discrimination is a very important problem in several fields.
- **Isometry invariant** shape discrimination has been approached with different tools, mostly via computation and comparison of **invariant signatures**, [HK03,Osada-02,Fro90,SC-00].
- The **Gromov-Hausdorff distance** (and certain variants) provides a **rigorous** and well motivated framework for studying shape matching under invariances [MS04,MS05,M07,M08].
- However, its direct computation leads to **NP hard** problems (BQAP: bottleneck quadratic assignment problems).

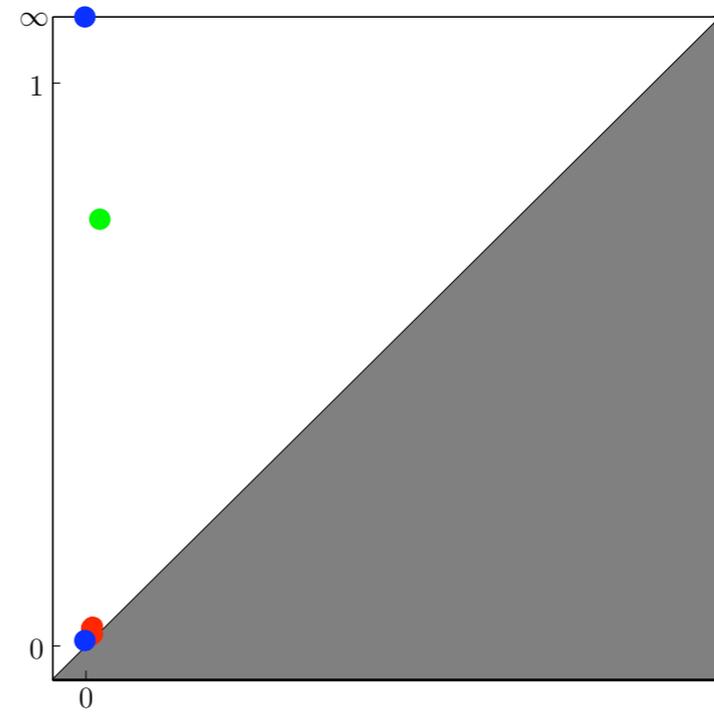
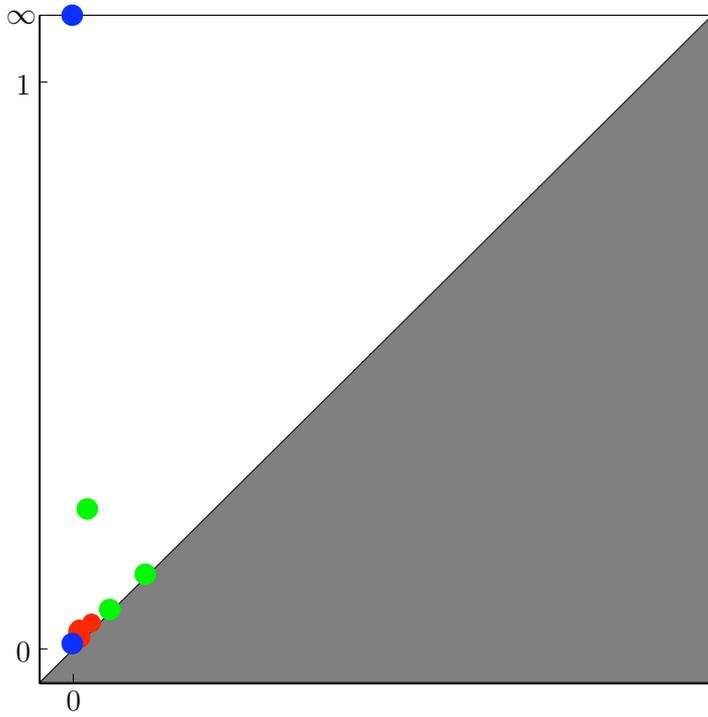
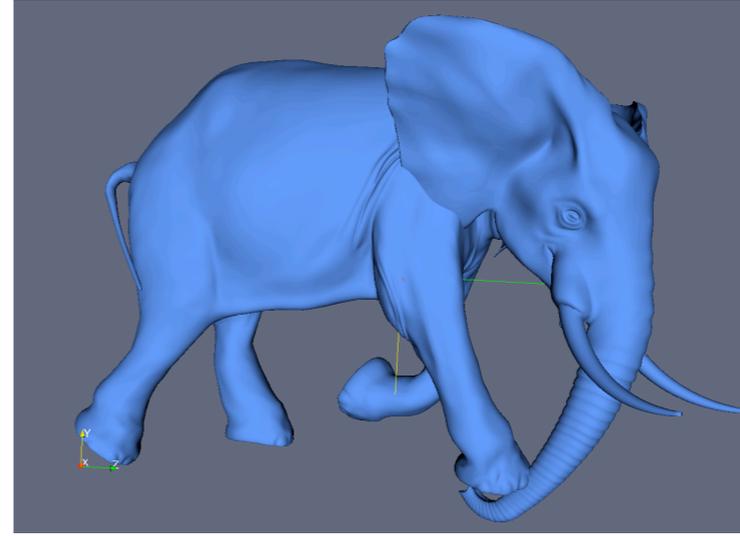
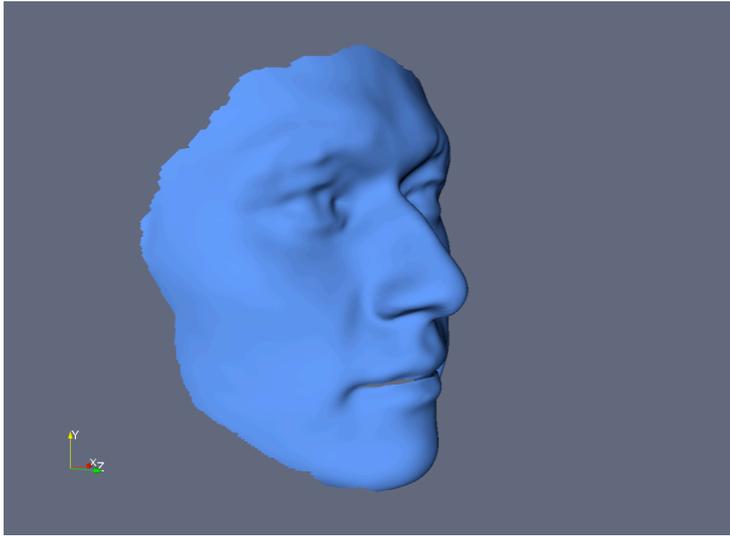


- Most of the effort has gone into finding **lower bounds** for the GH distance that use informative invariant signatures and lead to easier optimization problems [M07,M08].
- Using **persistent topology** [ELZ00], we obtain a new family of signatures and prove that they are **stable** w.r.t the GH distance: i.e., **we obtain lower bounds for the GH distance!**
- These lower bounds:
 - perform very well in practical application of shape discrimination.
 - lead to **BAPs** (bottleneck assignment problems) which can be solved in **polynomial time**.

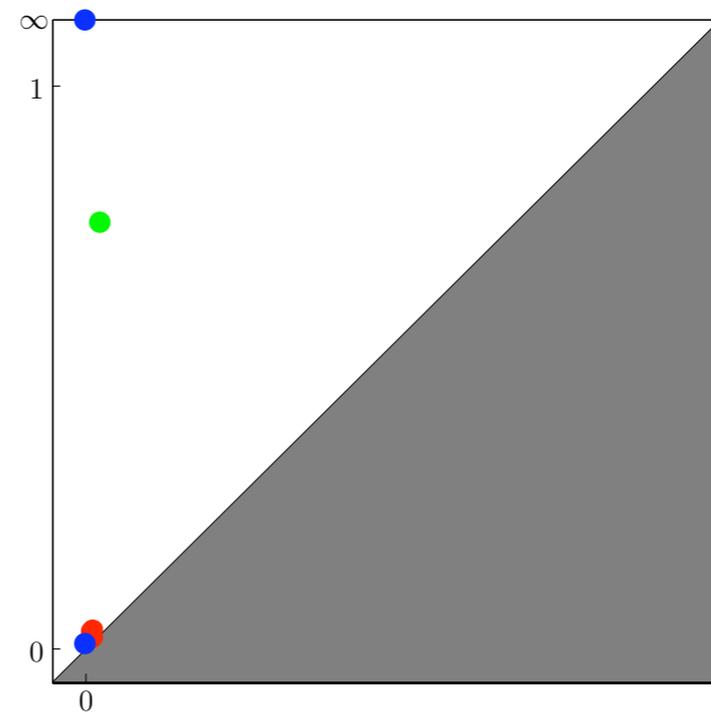
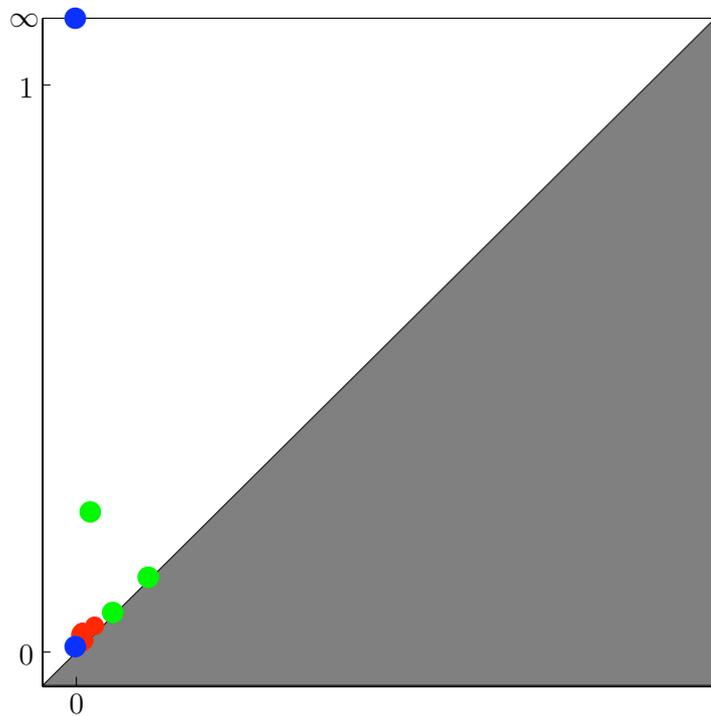
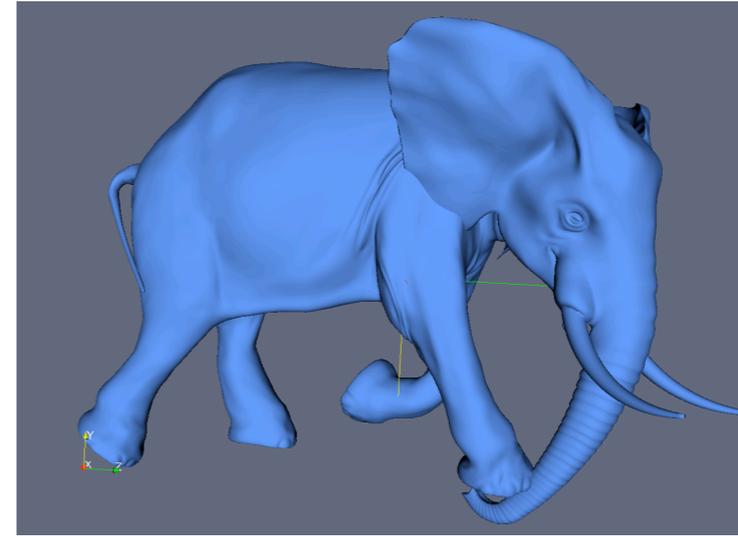
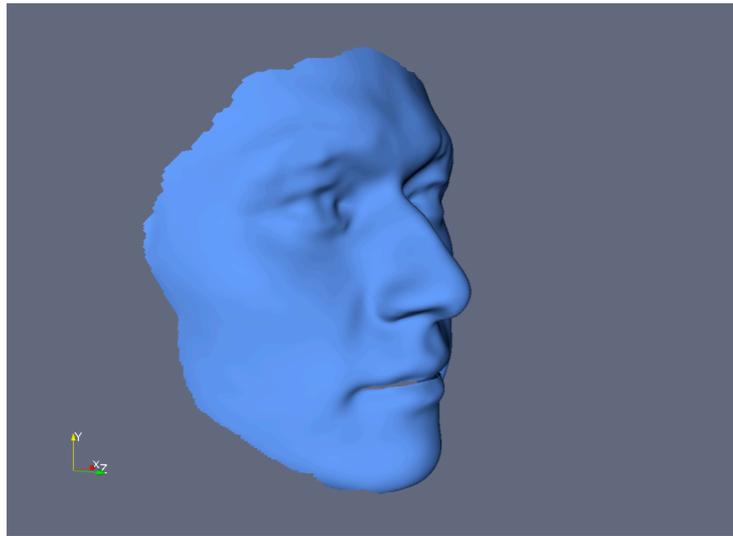
visual summary



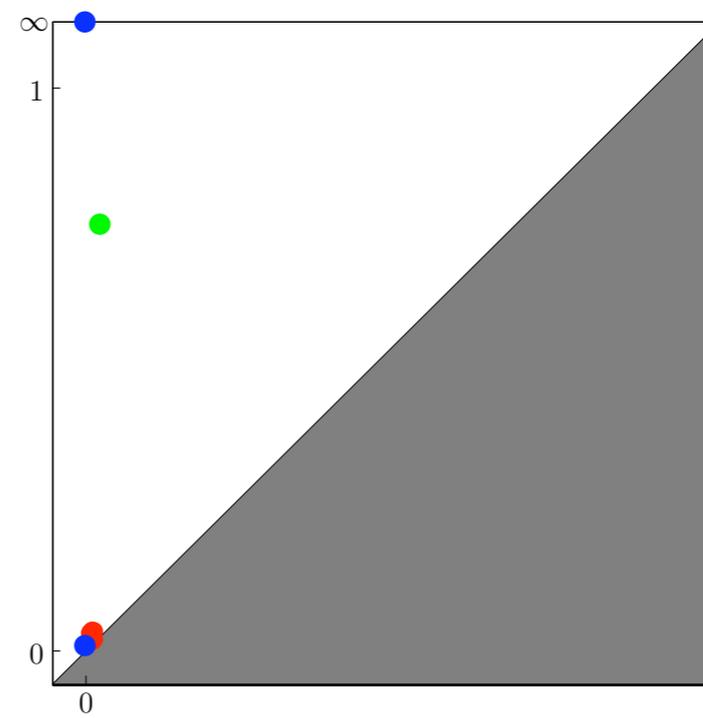
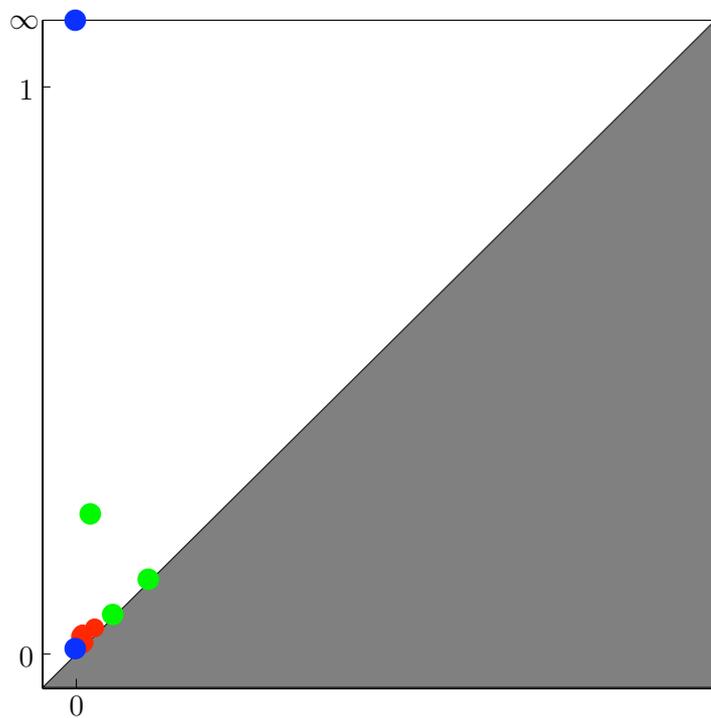
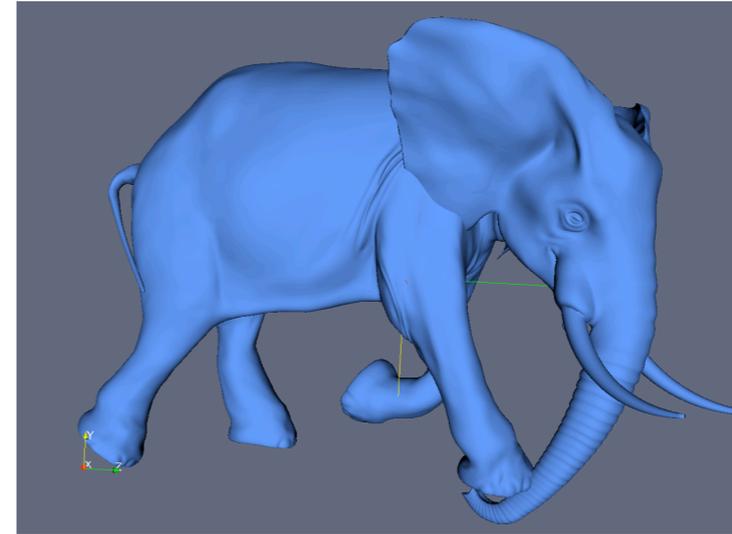
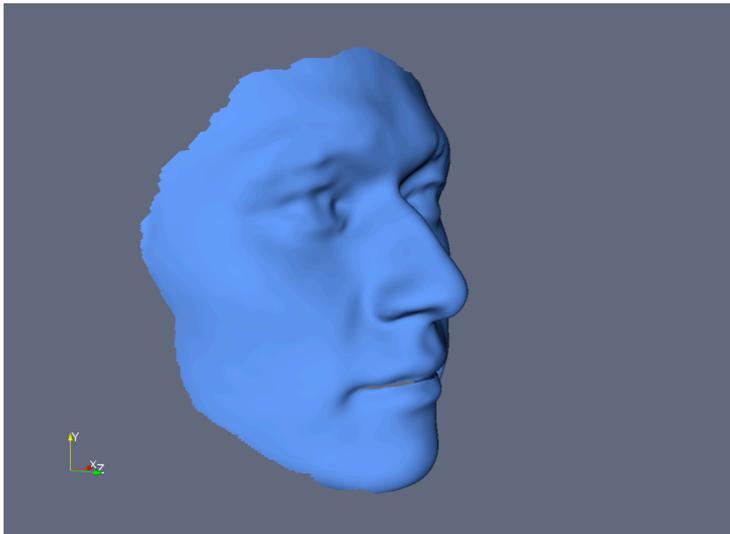
visual summary



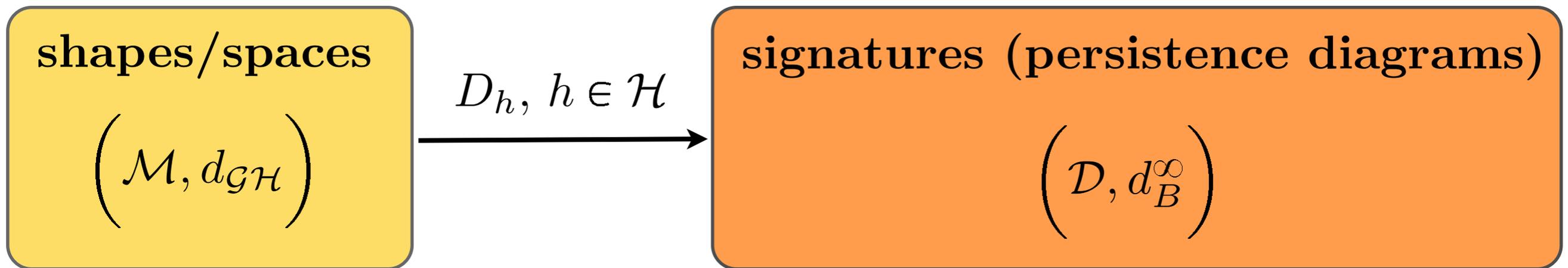
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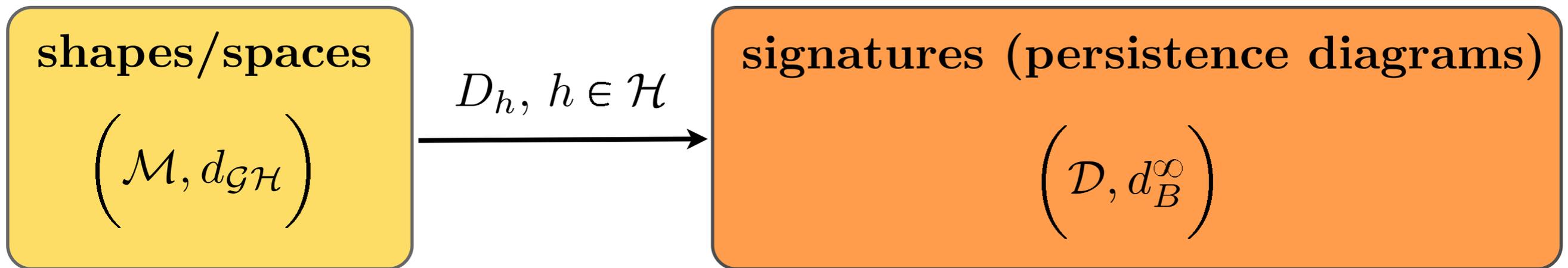
visual summary



- \mathcal{M} : collection of all shapes (finite metric spaces).
- \mathcal{D} : collection of all signatures (persistence diagrams).

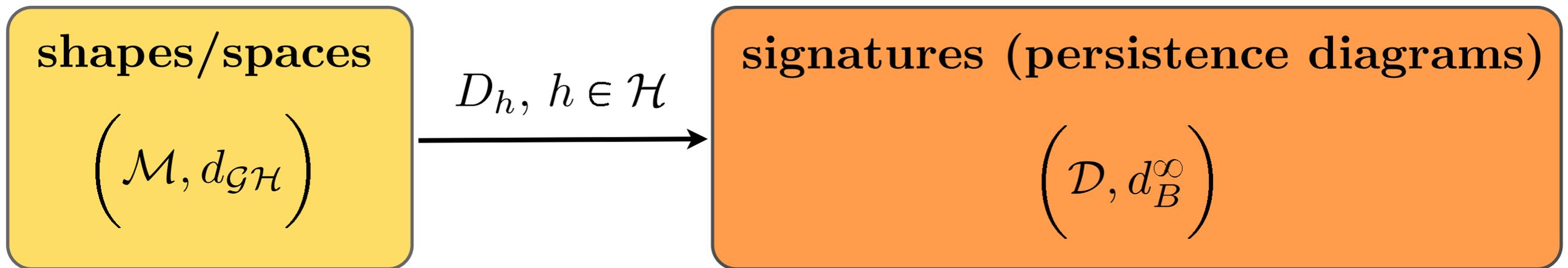


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X, Y

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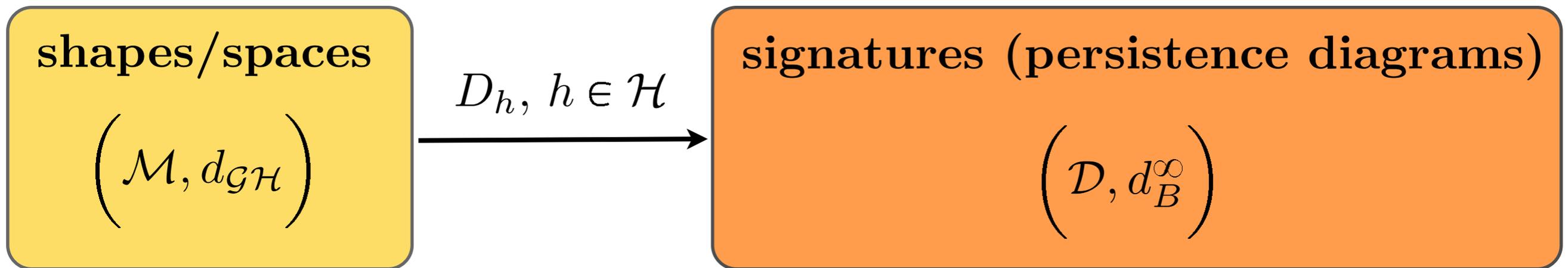


X, Y

$D_h(X), D_h(Y)$

$h \in \mathcal{H}$

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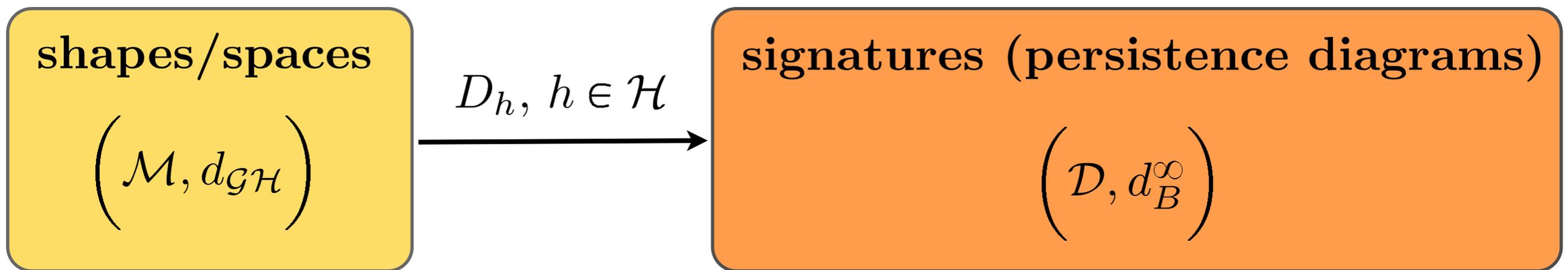
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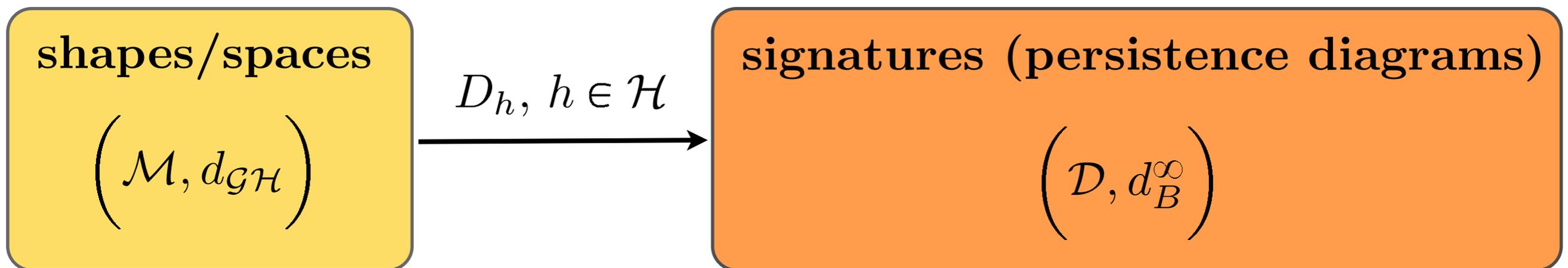
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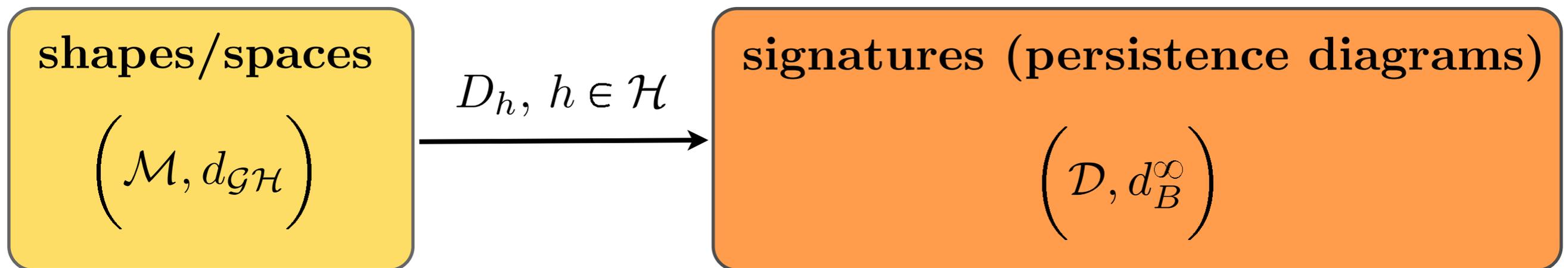
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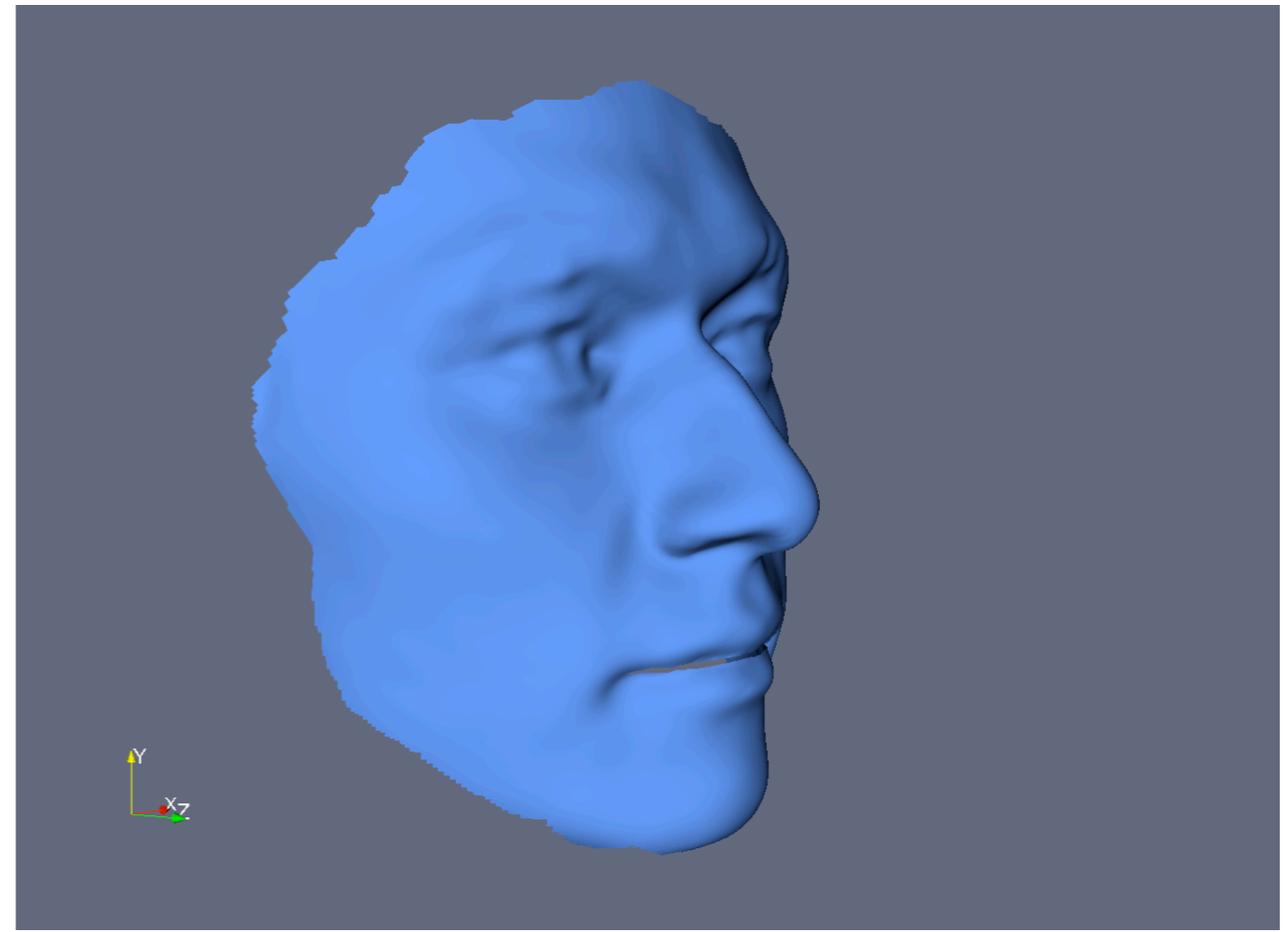
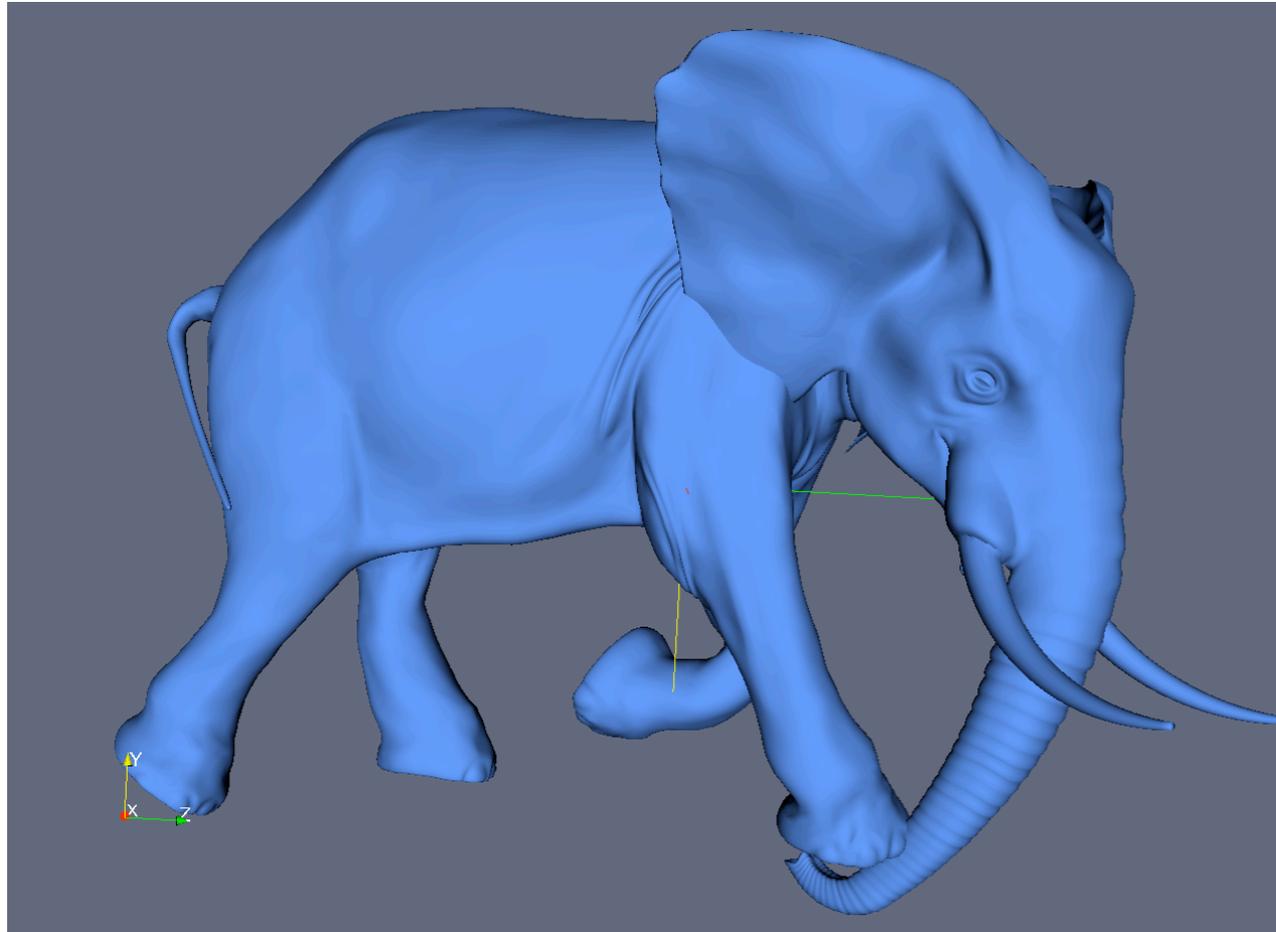
X, Y

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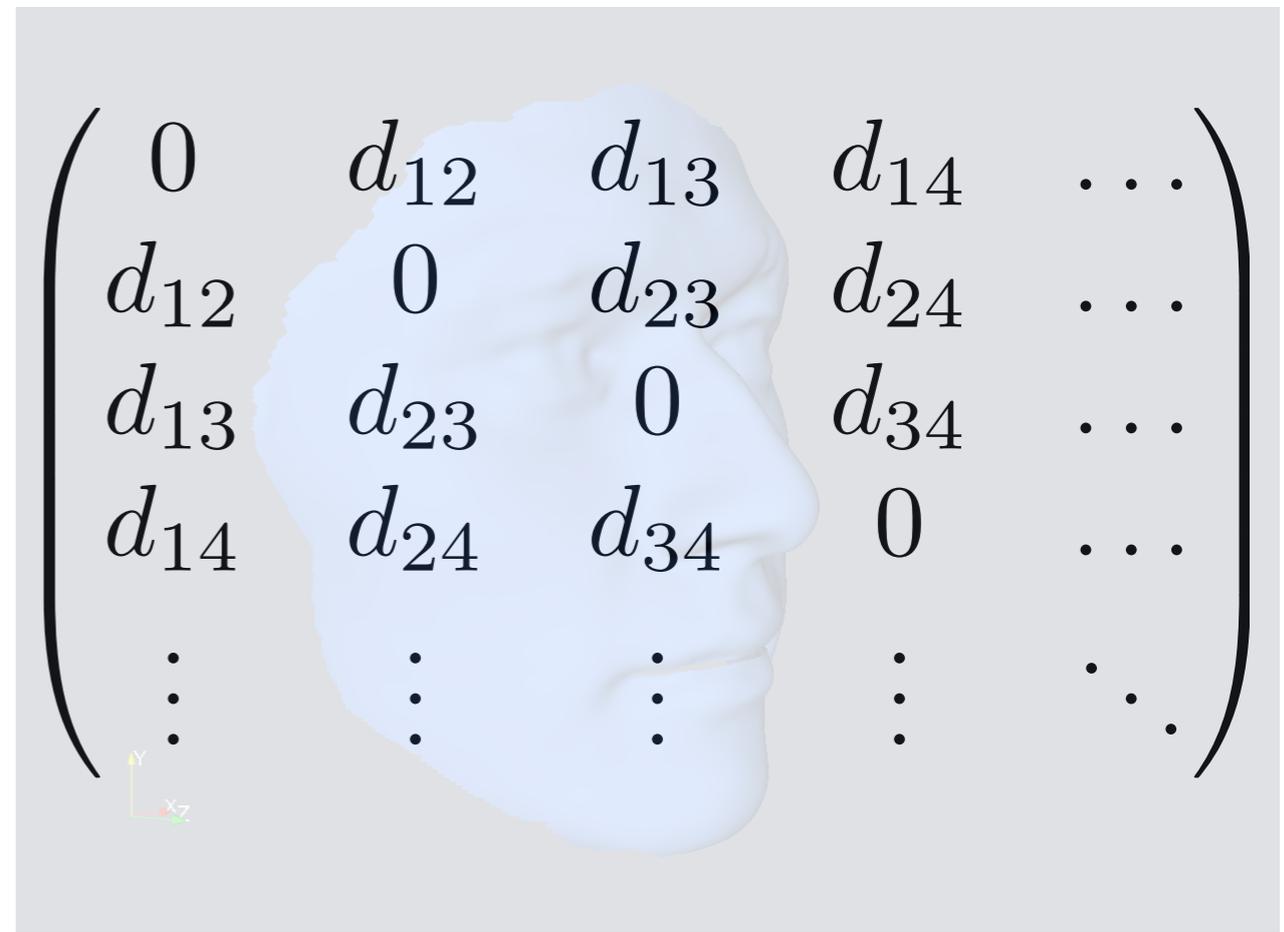
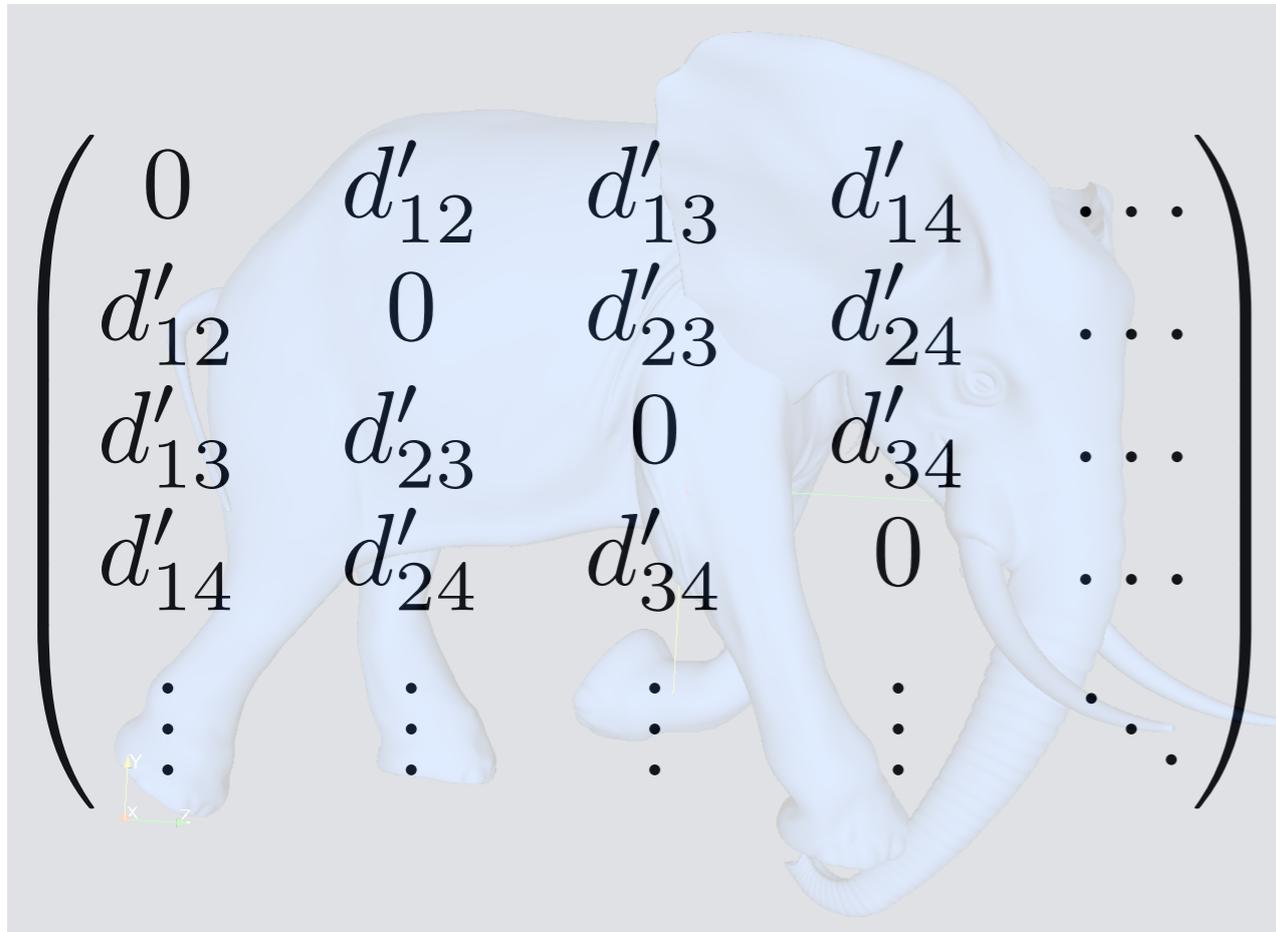
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$$d_{\mathcal{GH}}(X, Y) \geq d_B^\infty(D_h(X), D_h(Y))$$

Shapes as metric spaces

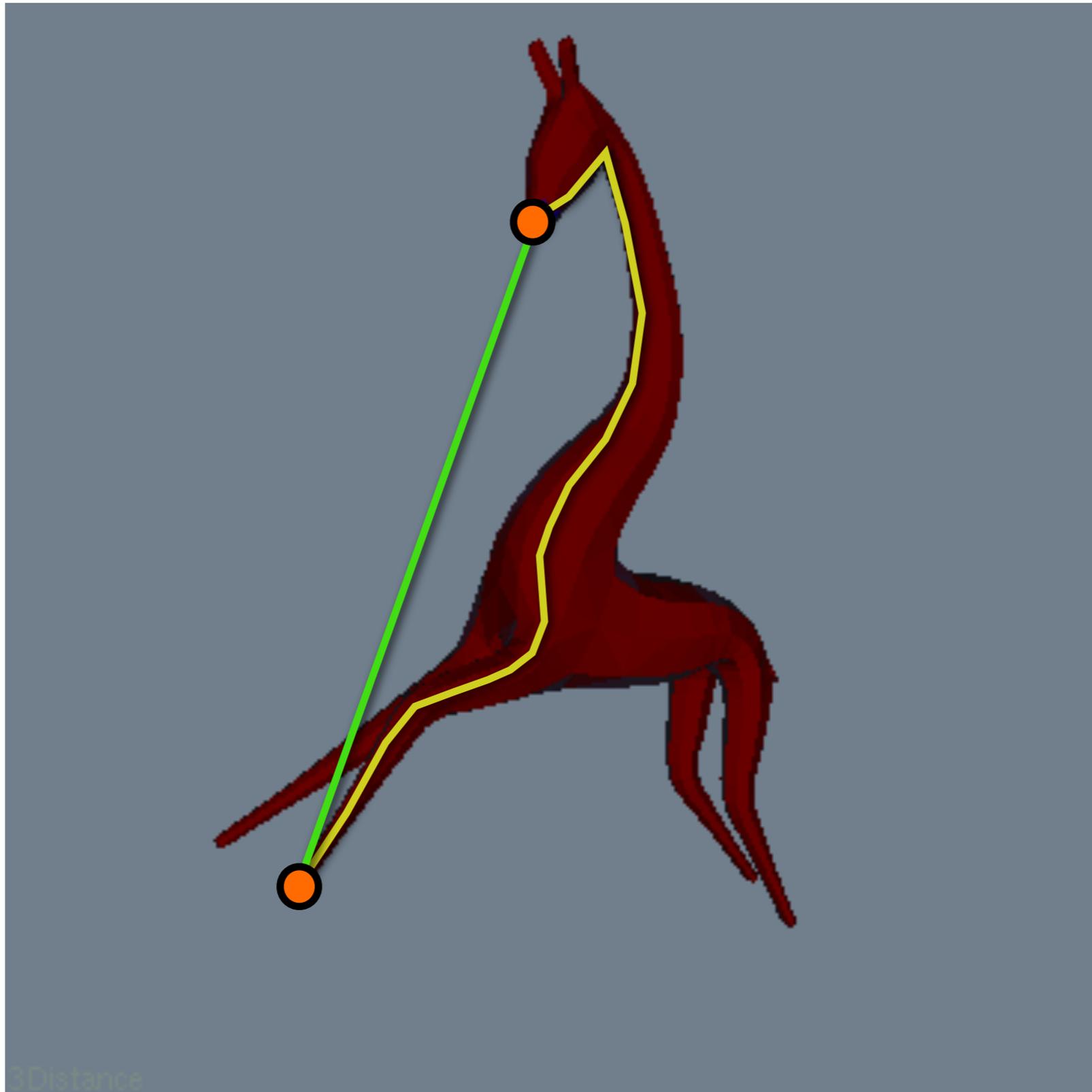


Shapes as metric spaces

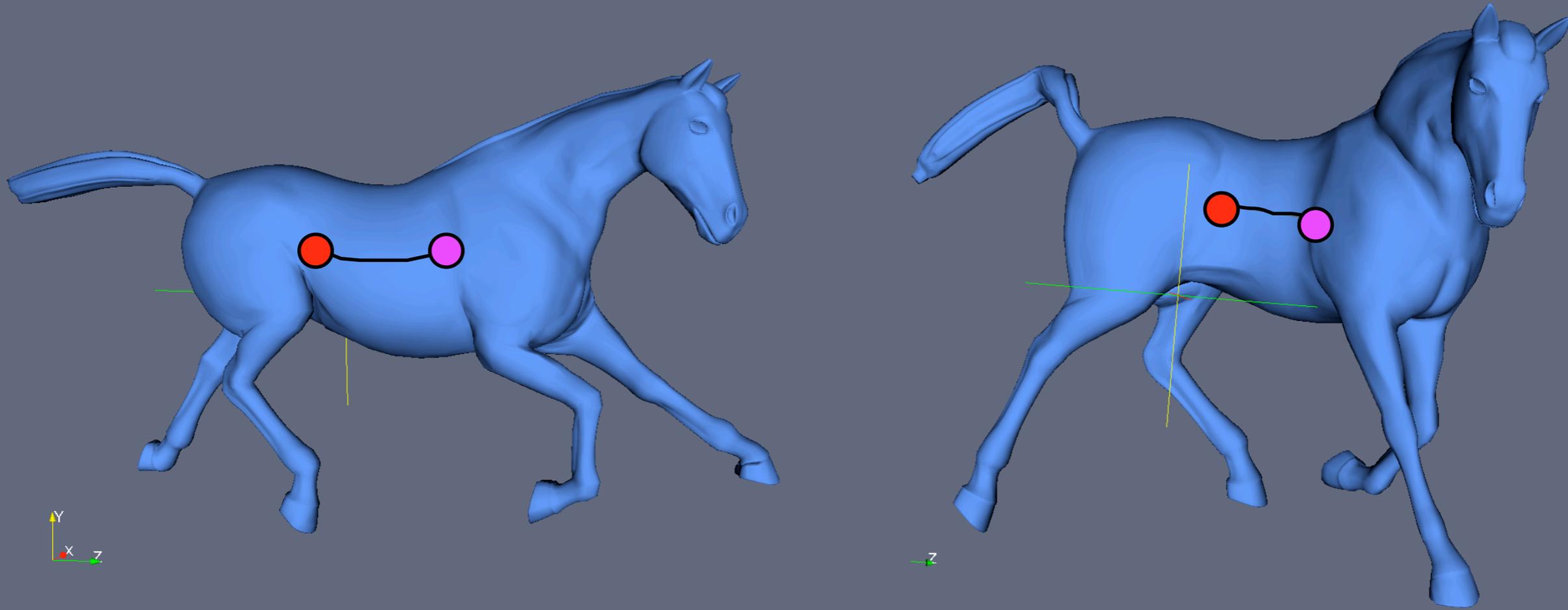


then use Gromov-Hausdorff distance..

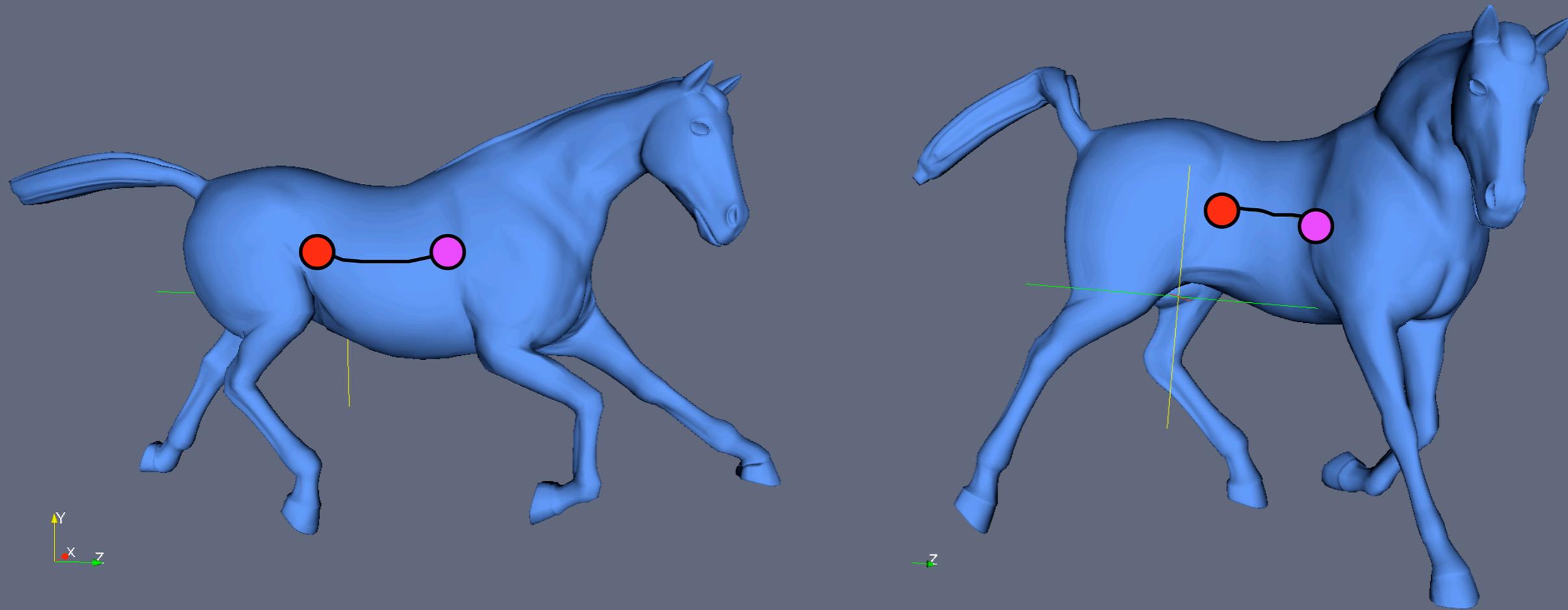
Choice of the metric: geodesic vs Euclidean



Invariance to isometric deformations (change in pose)

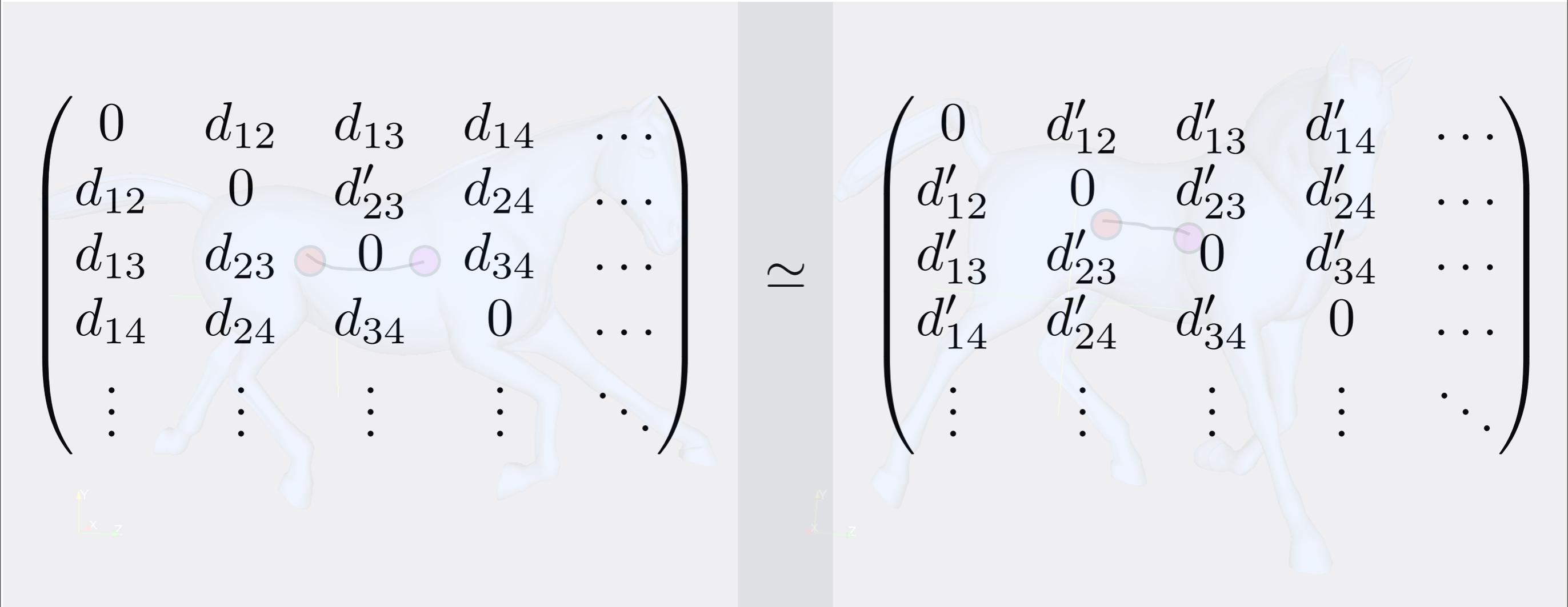


Invariance to isometric deformations (change in pose)



geodesic distance remains approximately constant

Invariance to isometric deformations (change in pose)


$$\begin{pmatrix} 0 & d_{12} & d_{13} & d_{14} & \dots \\ d_{12} & 0 & d'_{23} & d_{24} & \dots \\ d_{13} & d_{23} & 0 & d_{34} & \dots \\ d_{14} & d_{24} & d_{34} & 0 & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix} \approx \begin{pmatrix} 0 & d'_{12} & d'_{13} & d'_{14} & \dots \\ d'_{12} & 0 & d'_{23} & d'_{24} & \dots \\ d'_{13} & d'_{23} & 0 & d'_{34} & \dots \\ d'_{14} & d'_{24} & d'_{34} & 0 & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$

geodesic distance remains approximately constant

Definition [Correspondences]

For finite sets A and B , a subset $C \subset A \times B$ is a *correspondence* (between A and B) if and only if

- $\forall a \in A$, there exists $b \in B$ s.t. $(a, b) \in R$
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Let $\mathcal{C}(A, B)$ denote all possible correspondences between sets A and B .

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B

A

0	1	1	0	0	1	1
1	1	0	1	0	1	1
1	0	1	0	1	1	0
0	0	0	0	0	0	0
1	0	1	1	0	1	0

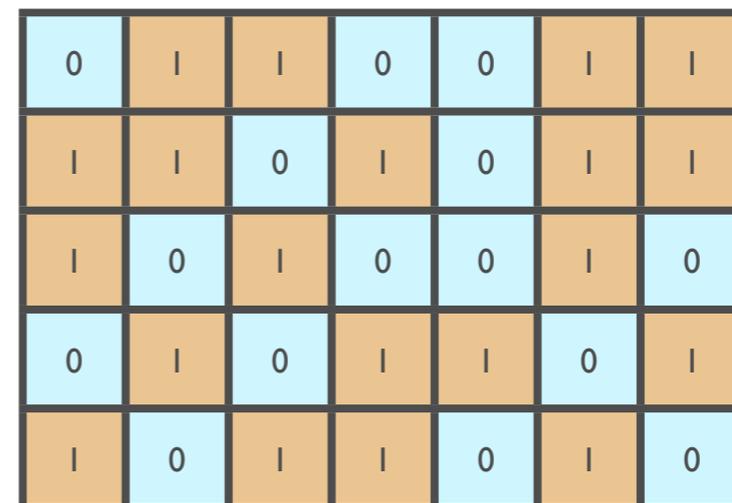
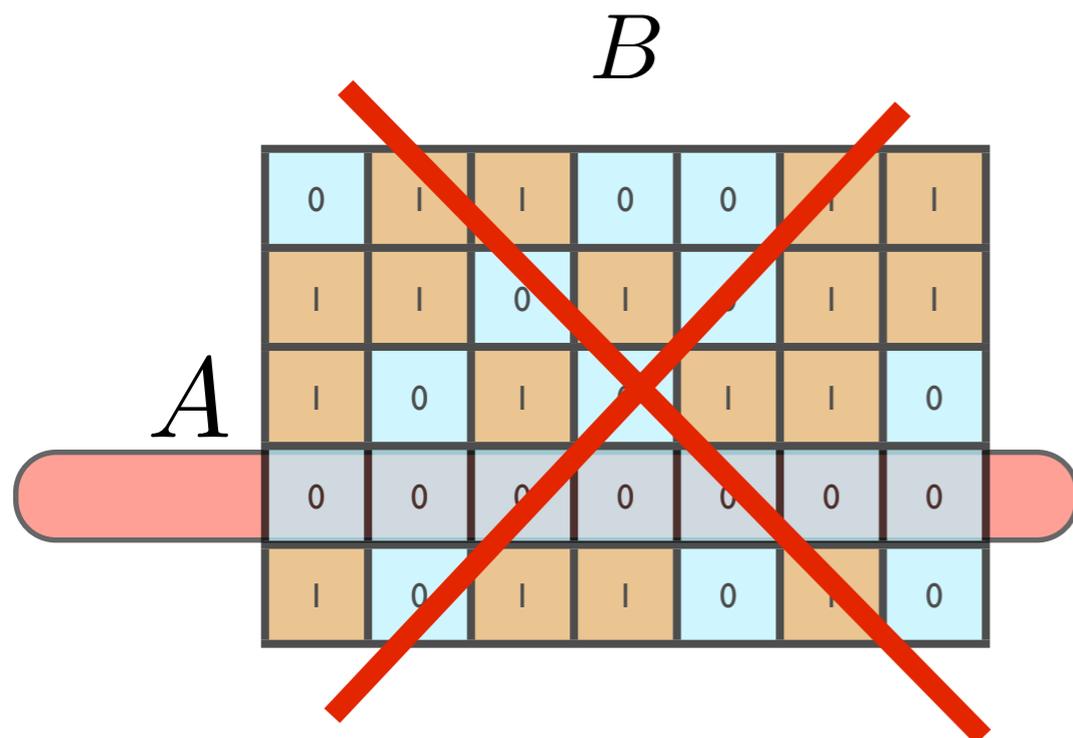
0	1	1	0	0	1	1
1	1	0	1	0	1	1
1	0	1	0	0	1	0
0	1	0	1	1	0	1
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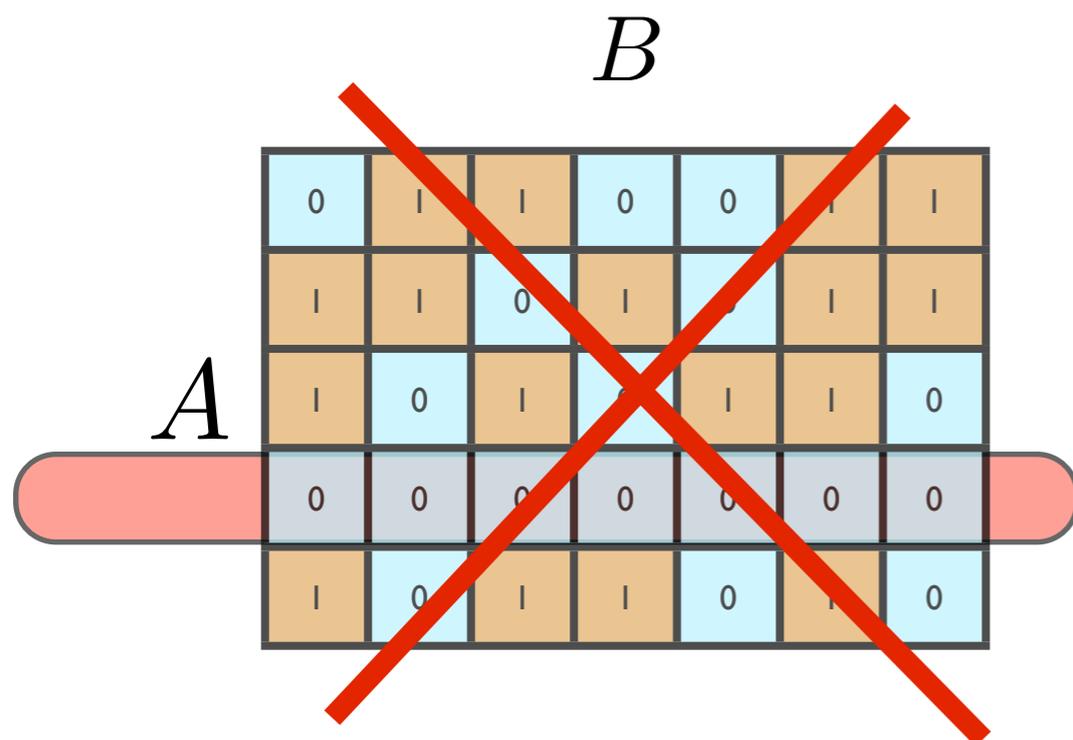


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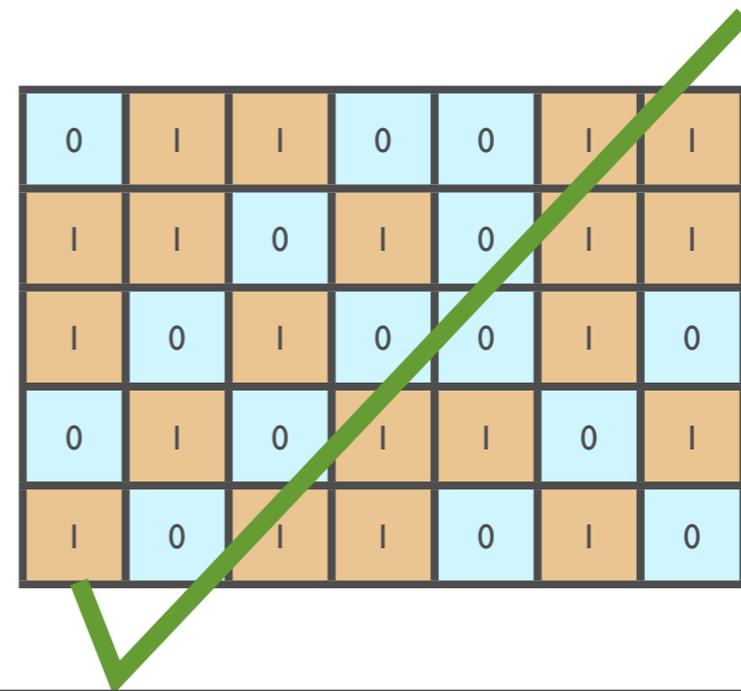
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9

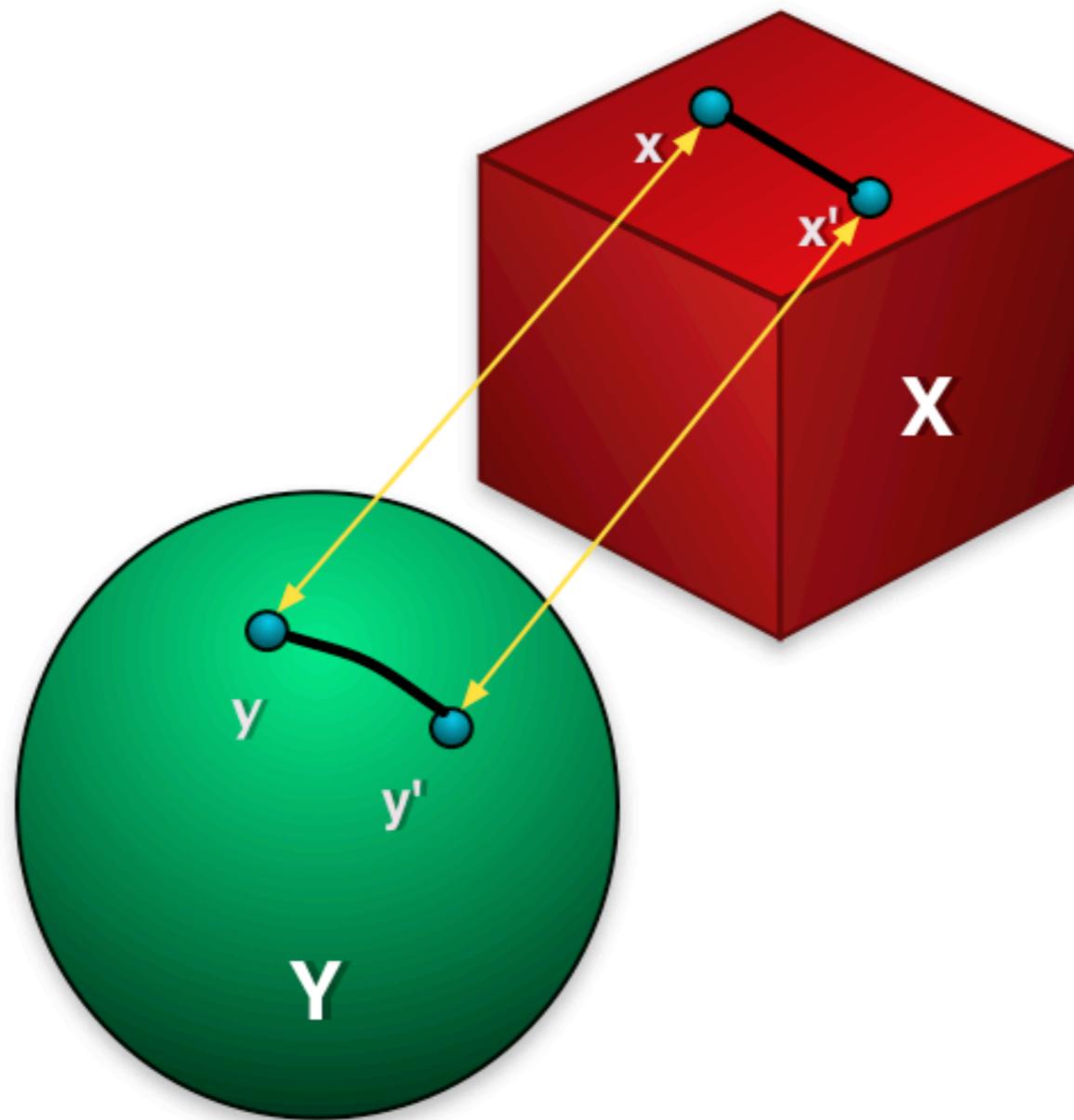


Definition. [BBI] For finite metric spaces (X, d_X) and (Y, d_Y) , define the *Gromov-Hausdorff distance* between them by

$$d_{\mathcal{GH}}(X, Y) = \frac{1}{2} \min_C \max_{(x, y), (x', y') \in C} |d_X(x, x') - d_Y(y, y')|$$

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Construction of our signatures

- Our signatures take the form of **persistence diagrams**: we capture certain topological and metric information from the shape.
- First example: construction based on **Rips filtrations**: Let (X, d_X) be a shape.

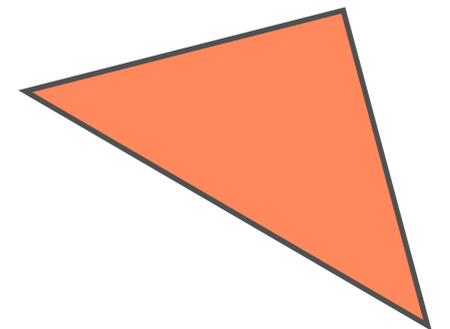
– Let $K_d(X)$ be the d -dimensional **full simplicial complex** on X .

– To each $\sigma = [x_0, x_1, \dots, x_k] \in K_d(X)$ assign its **filtration time**

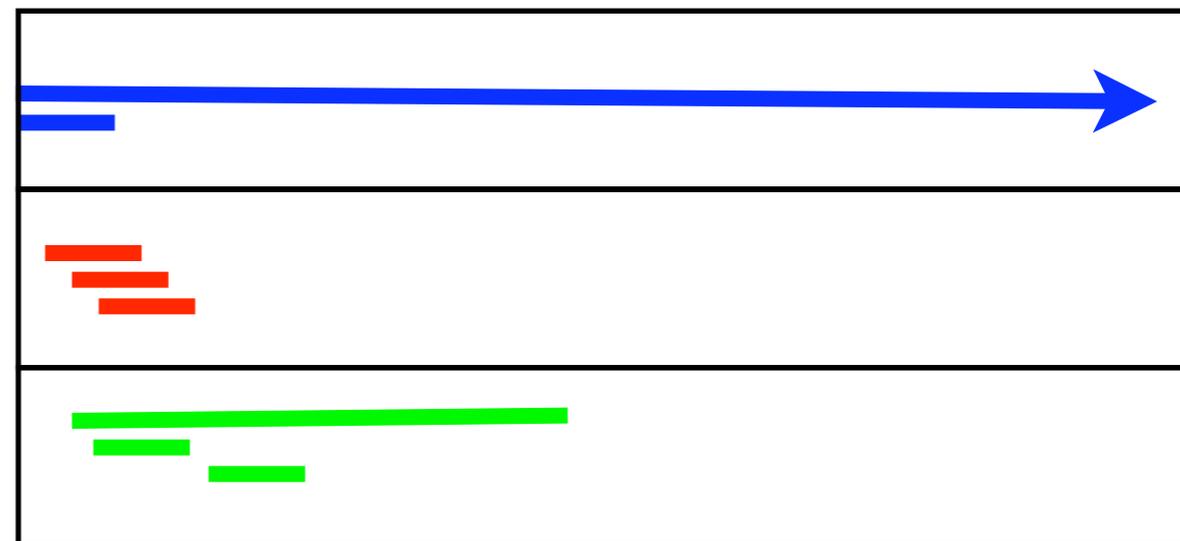
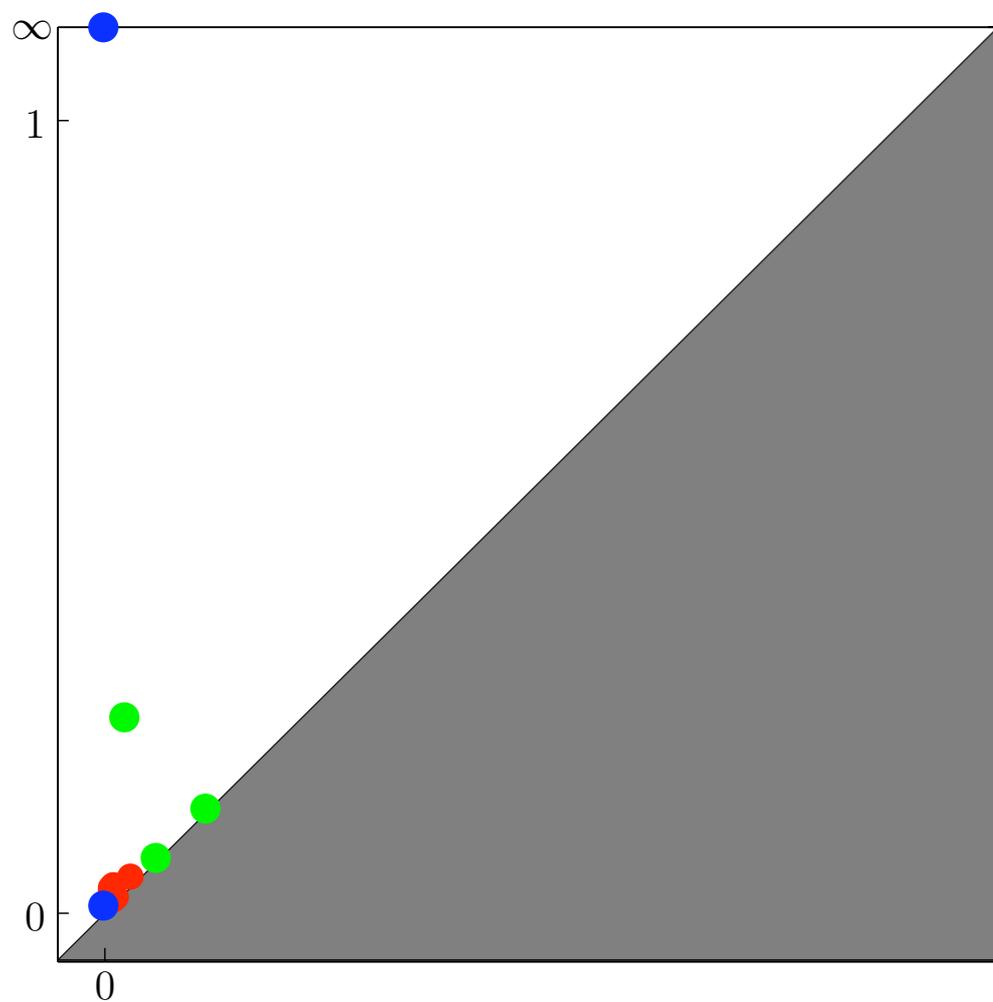
$$F(\sigma) := \frac{1}{2} \max_{i,j} d_X(x_i, x_j)$$

– This gives rise to a **filtration** $(K_d(X), F)$.

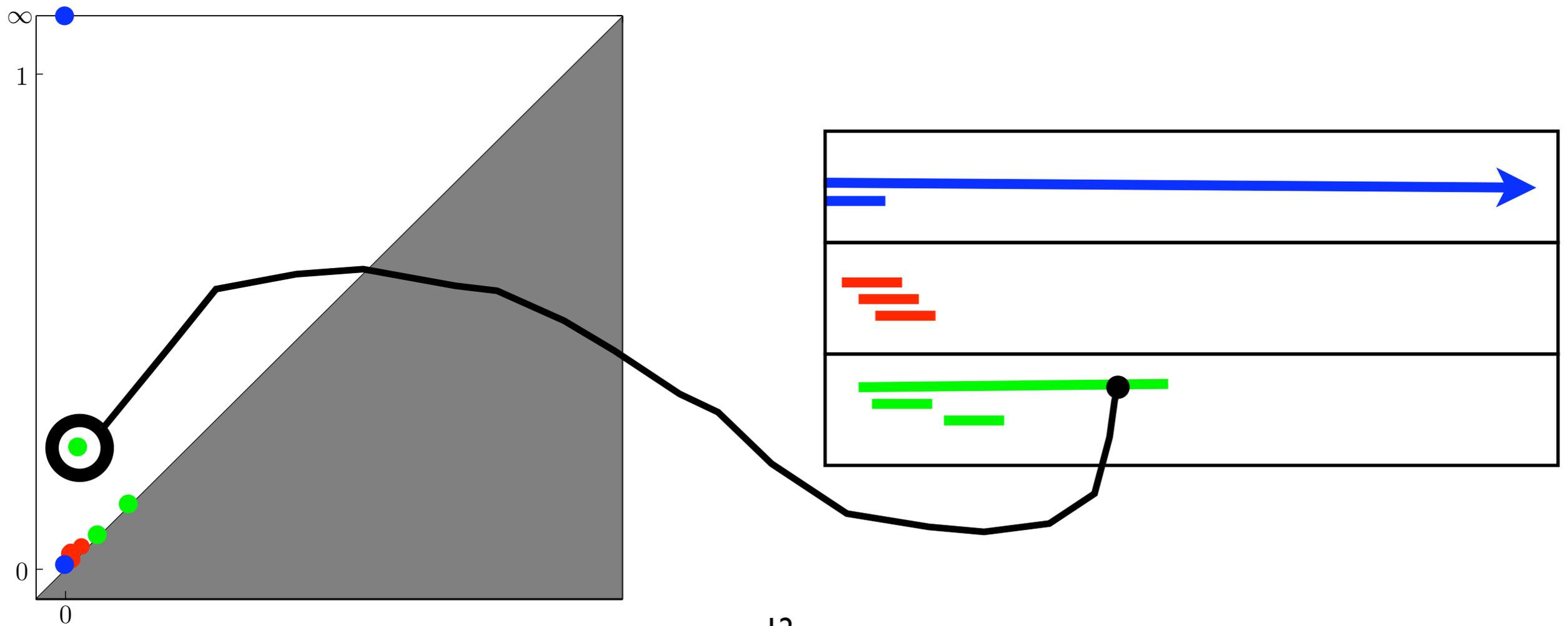
– Apply **persistence algorithm** [ELZ00] to summarize topological information in the filtration and obtain **persistence diagram**.



- Persistence diagrams are **colored multi-subsets** of the extended real plane.. can also be represented as **barcodes**.
- Let \mathcal{D} denote the collection of all persistence diagrams. Compare two different persistence diagrams with **bottleneck distance** \implies view $(\mathcal{D}, d_B^\infty)$ as a metric space.

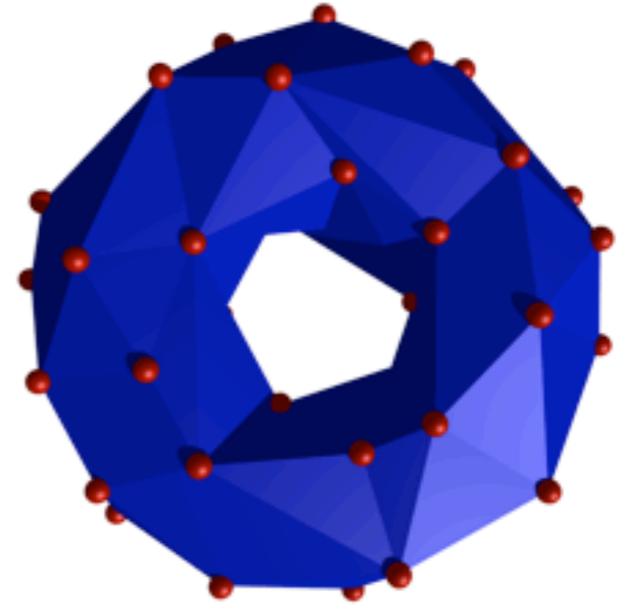
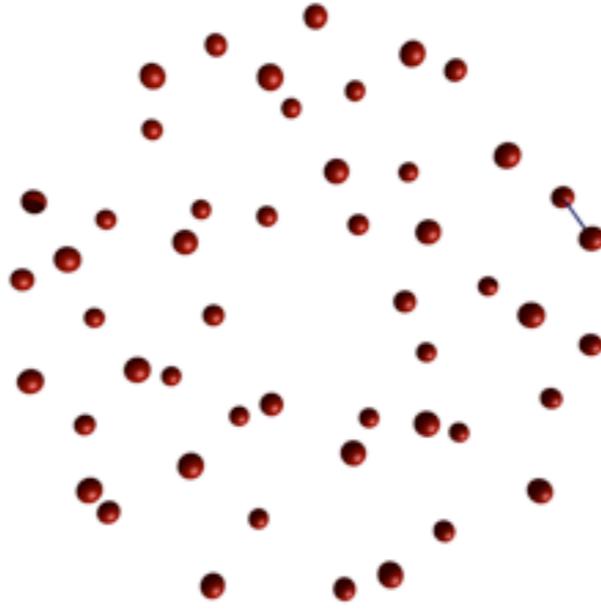
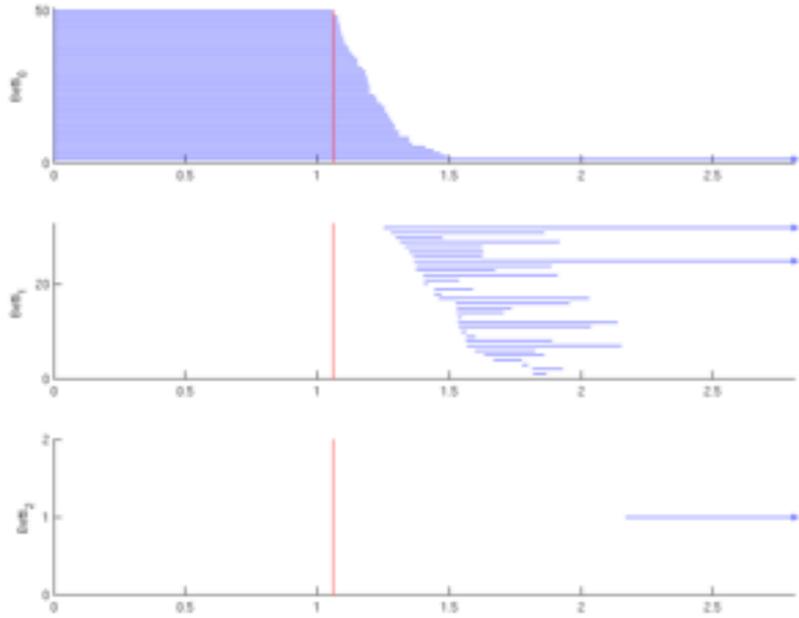


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Example: Rips filtration on a torus

Example: Rips filtration on a torus



Our signatures: more richness

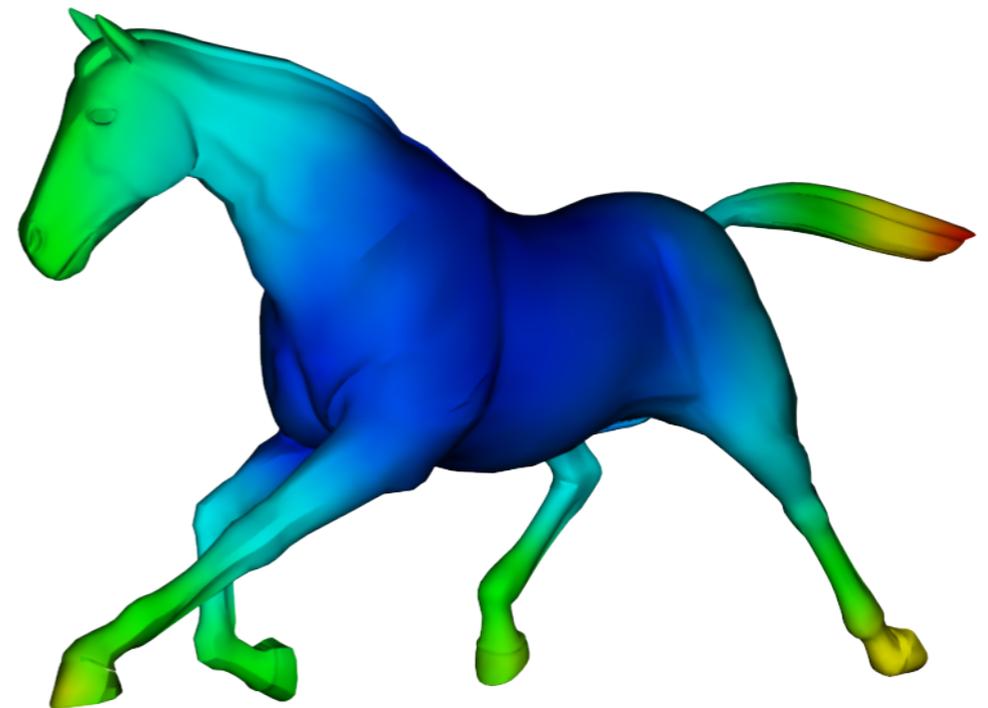
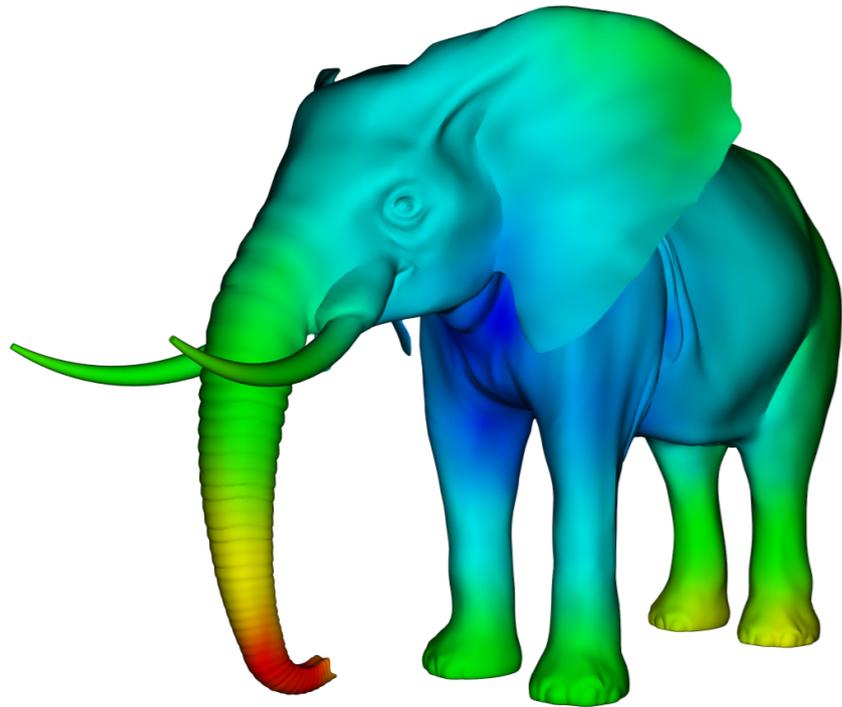
- Let's assume that in there is also a function defined on the shape: (X, d_X, f_X) . Then, we redefine the filtration values of $\sigma = [x_0, x_1, \dots, x_k]$

$$F(\sigma) = \max \left(\frac{1}{2} \max_{i,j} d_X(x_i, x_j), \max_i f_X(x_i) \right)$$

- Again, this gives rise to a **filtration**: $(K_d(X), F) \implies$ use **persistence algorithm** to obtain a persistence diagram.
- This increases discrimination power!
- We denote by \mathcal{H} a family of maps that attach a function to a given finite metric space.
- Then, for each $h \in \mathcal{H}$, we denote by $D_h(X)$ the persistence diagram arising from the filtration above. This constitutes our family projection onto \mathcal{D} .

Example (Eccentricity). *To each finite metric space (X, d_X) one can assign the **eccentricity** function:*

$$ecc_X(x) = \max_{x' \in X} d_X(x, x').$$



$$h \in \mathcal{H}$$

shapes/spaces

$$\left(\mathcal{M}, d_{\mathcal{GH}} \right)$$



signatures (persistence diagrams)

$$\left(\mathcal{D}, d_B^\infty \right)$$

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$$X, Y$$

$$D_h(X), D_h(Y)$$

$$\cancel{d_{\mathcal{GH}}(X, Y)}$$

$$d_B^\infty(D_h(X), D_h(Y))$$

Theorem (stability of our signatures). *For all $X, Y \in \mathcal{M}$,*

$$d_{\mathcal{GH}}(X, Y) \geq C(h) \cdot d_B^\infty(D_h(X), D_h(Y)).$$

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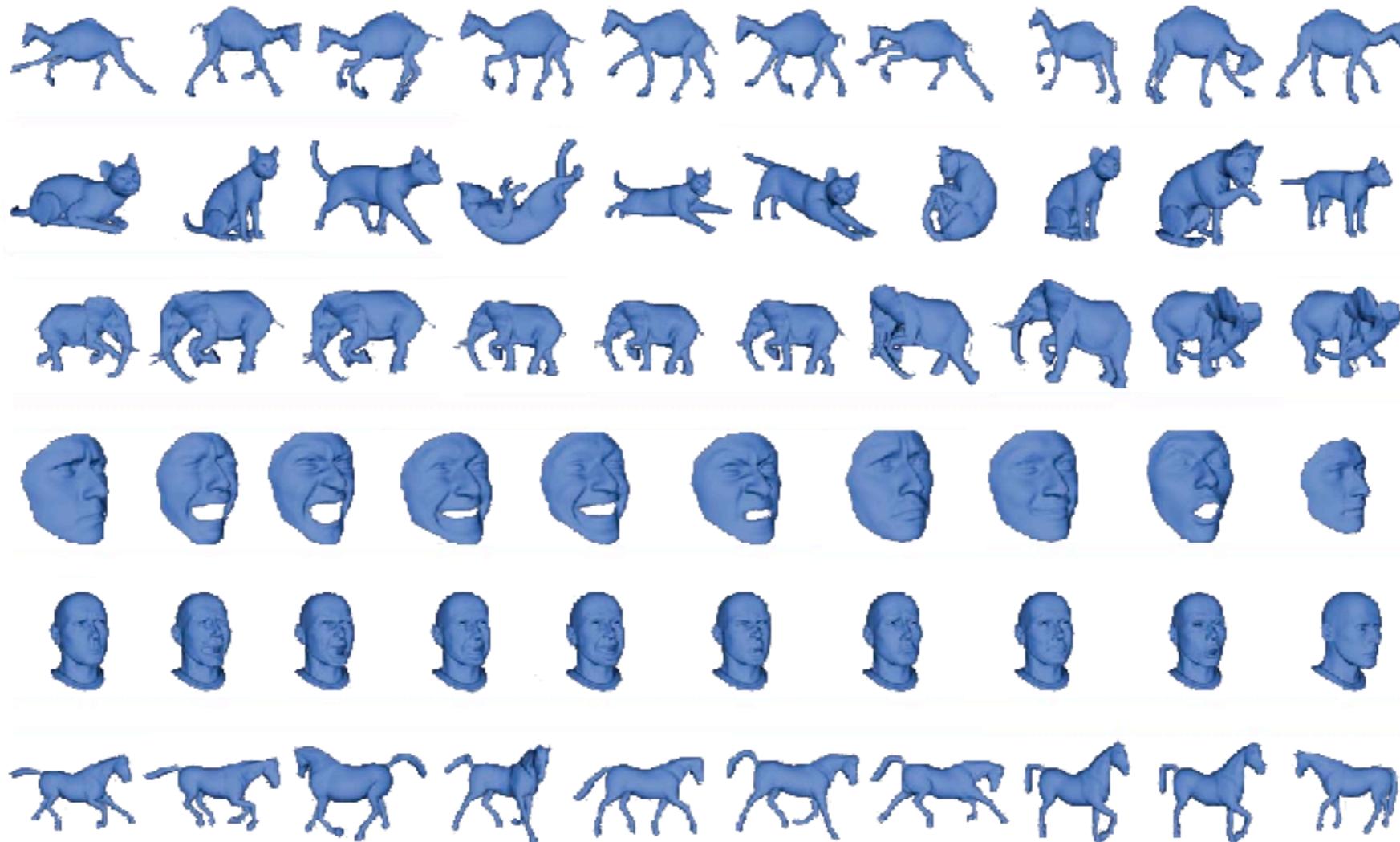
$$d_{\mathcal{GH}}(X, Y) \geq \sup_{h \in \mathcal{H}} C(h) \cdot d_B^\infty(D_h(X), D_h(Y)).$$

Remark.

- *Proof relies on properties of the GH distance and new results on the stability of persistence diagrams [CCGG009].*
- *For a given h , the computation leads to a **BAP** which can be solved in polynomial time.*
- *There are adaptations one can do in practice to speed up, see paper.*
- *One can obtain more generality and discrimination power by working in the class of **mm-spaces**: shapes are represented as triples (X, d_X, μ_X) where μ_X are **weights** assigned to each point see [M07] and paper.*
- *Our results include **stability of Rips persistence diagrams**.*

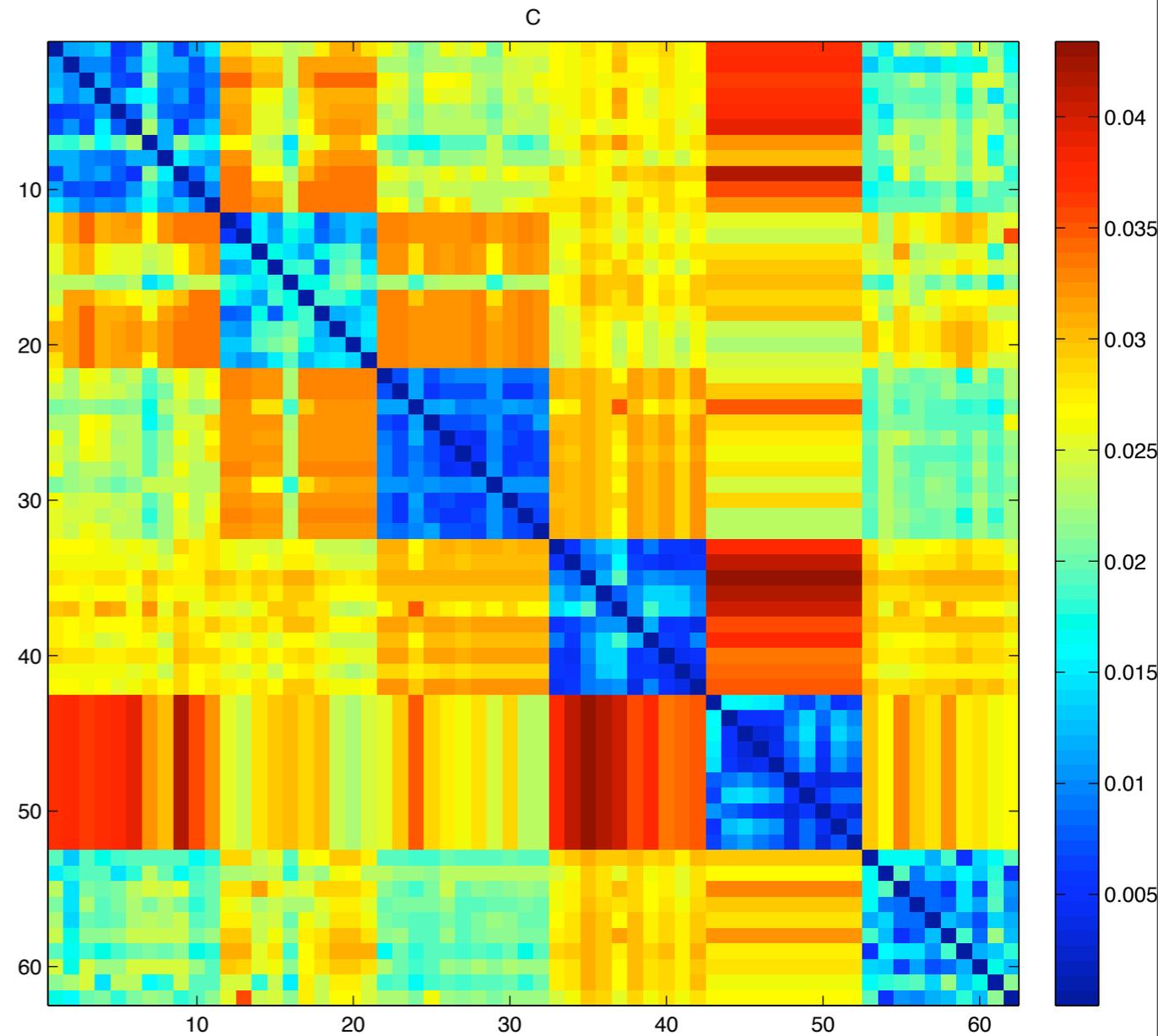
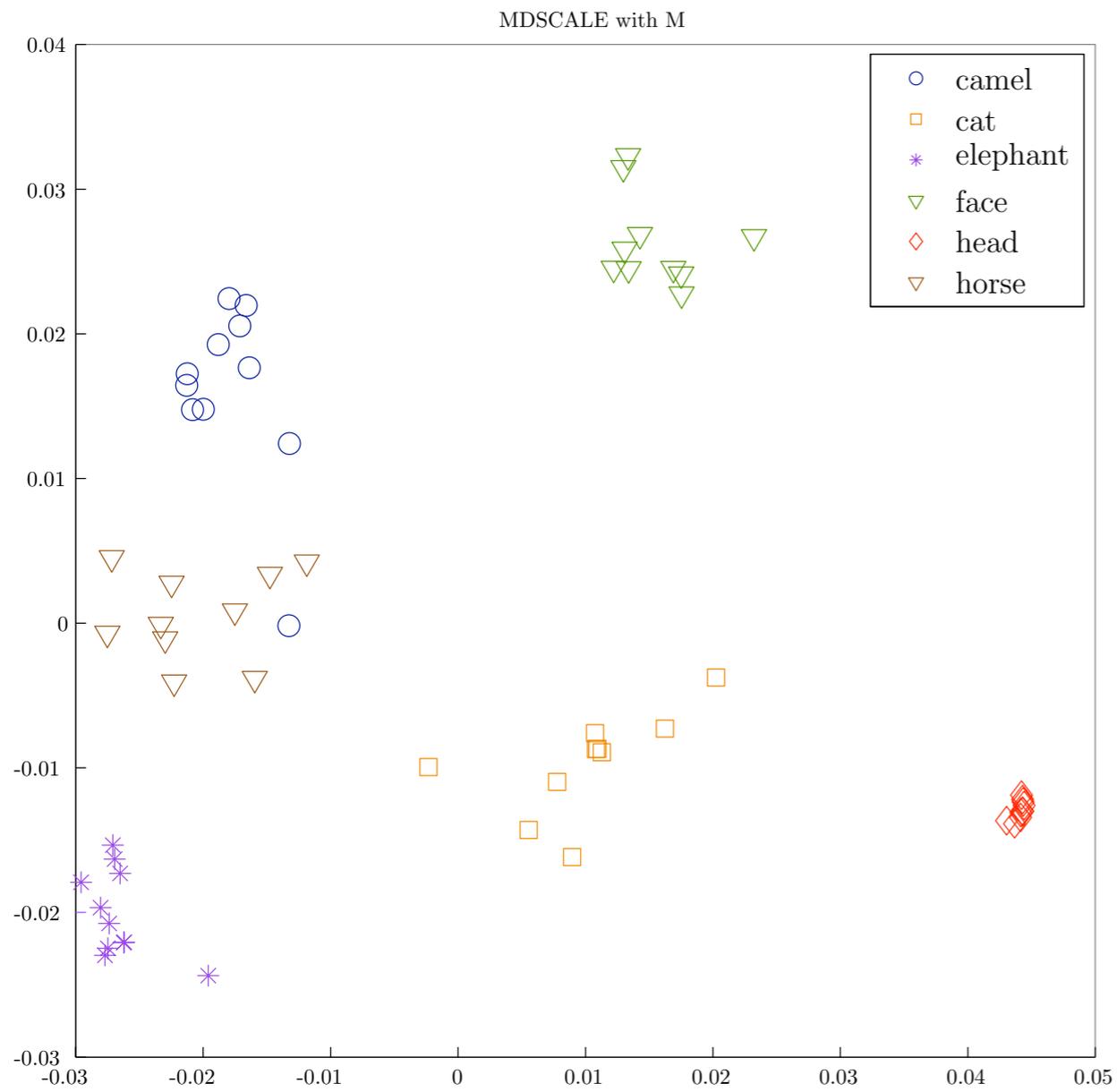
Some experiments

- Summer database: 62 shapes total, 6 classes. Used graph estimate of the geodesic distance. Number of vertices ranged from 7K to 30K.



Some experiments

- Sumner database: 62 shapes total, 6 classes. Used graph estimate of the geodesic distance. Number of vertices ranged from 7K to 30K.
- Subsampled shapes and retained subsets of 300 points (farthest point sampling). Normalized distance matrices.
- Used the mm-space representation of shapes: weights were based on Voronoi regions.
- Used several functions $\pm\lambda \cdot h$ for λ in a finite subset of scales.
- Obtained 4% (or 2%) classification error in a 1-nn classification problem.



Discussion

- **Summary of our proposal:**

- Use the metric (or mm-space) representation of shapes.
- Formulate the shape matching problem using the Gromov-Hausdorff distance.
- Compute our signatures for shapes.
- Solve the BAP lower bounds: computationally easy! By our theorem, the computed quantities give lower bounds for the GH distance.

- **Implications and Future directions:**

- We do not need a mesh— general: can be applied to any dataset.
- We obtain stability of Rips persistence diagrams.
- Richness of the family \mathcal{H} ? how close can I get to the GH distance?
- Local signatures: more discrimination.
- Extension to partial shape matching: which (local) signatures are useful for this?

Acknowledgements

- ONR through grant N00014-09-1-0783
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- INRIA-Stanford associated TGDA team.

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Let $X_1, X_2 \subset Z$ be two different samples of the same shape Z , and Y another shape then

$$|d_{\mathcal{GH}}(X_1, Y) - d_{\mathcal{GH}}(X_2, Y)| \leq d_{\mathcal{GH}}(X_1, X_2) \leq r_1 + r_2$$