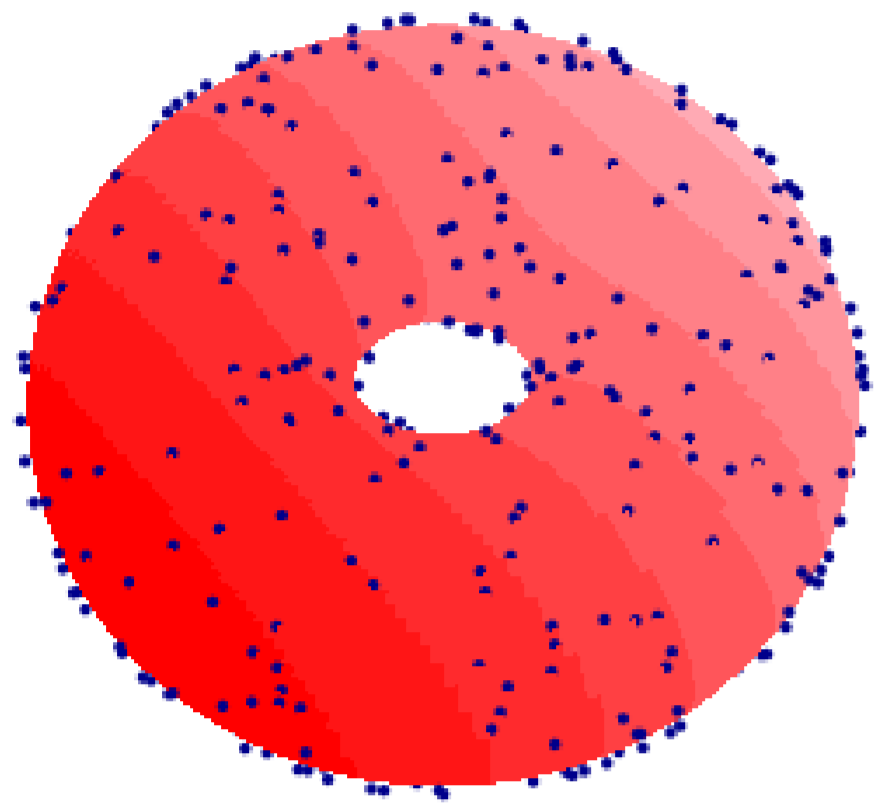


Simulation of PCD* on parametrized manifolds

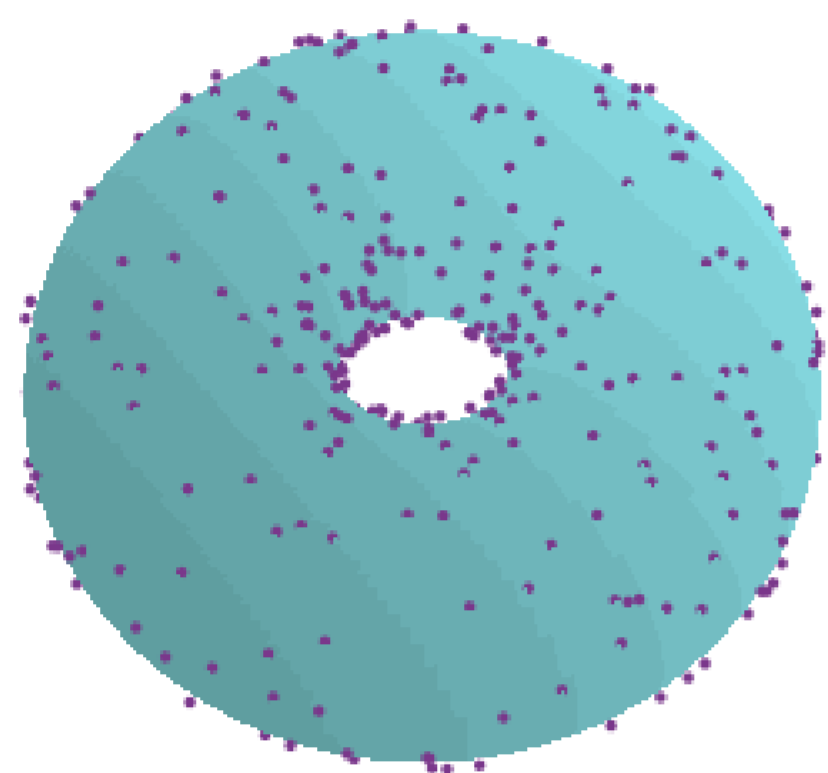
Simulation methods

How to sprinkle a donut?

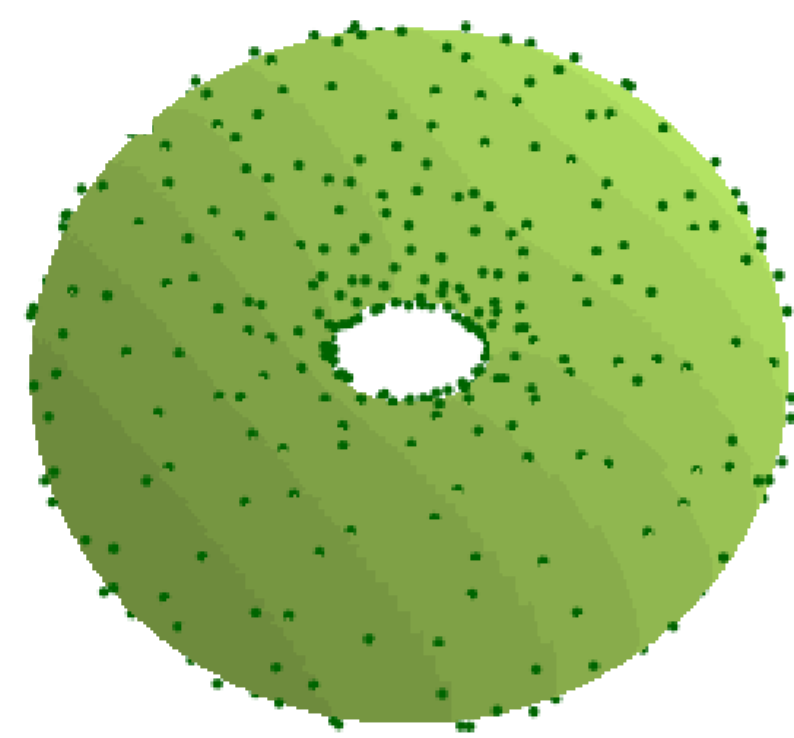


Useful for testing TDA algorithms

How to sample "uniformly" from a (parametrized) manifold M ?



How does the distribution matter?



Method by Diaconis et al. [2]

Sprinkling uniformly

1

Based on Area Formula:

M parametrized by $f, x = f(\theta)$

$$\int_A g(f(x)) J_m f(x) \lambda^m(dx)$$

$$= \int_{\mathbb{R}^n} g(y) N(f|_A, y) \mathcal{H}^m(dy)$$

Lebesgue measure $\lambda^m(dx)$ is transformed via the m-dimensional Jacobian of f to the Hausdorff measure $\mathcal{H}^m(dy)$ on the manifold. The term $N(f|_A, y)$ represents the cardinality of the fiber of y .

If $g = \frac{1}{\mathcal{H}^m(M)}$ we can sample \mathcal{H}^m -uniformly, sampling from $\frac{J_m f}{\mathcal{H}^m(M)}$ on the domain of f .

Torus is only example in [2], already implemented in TDA R library.

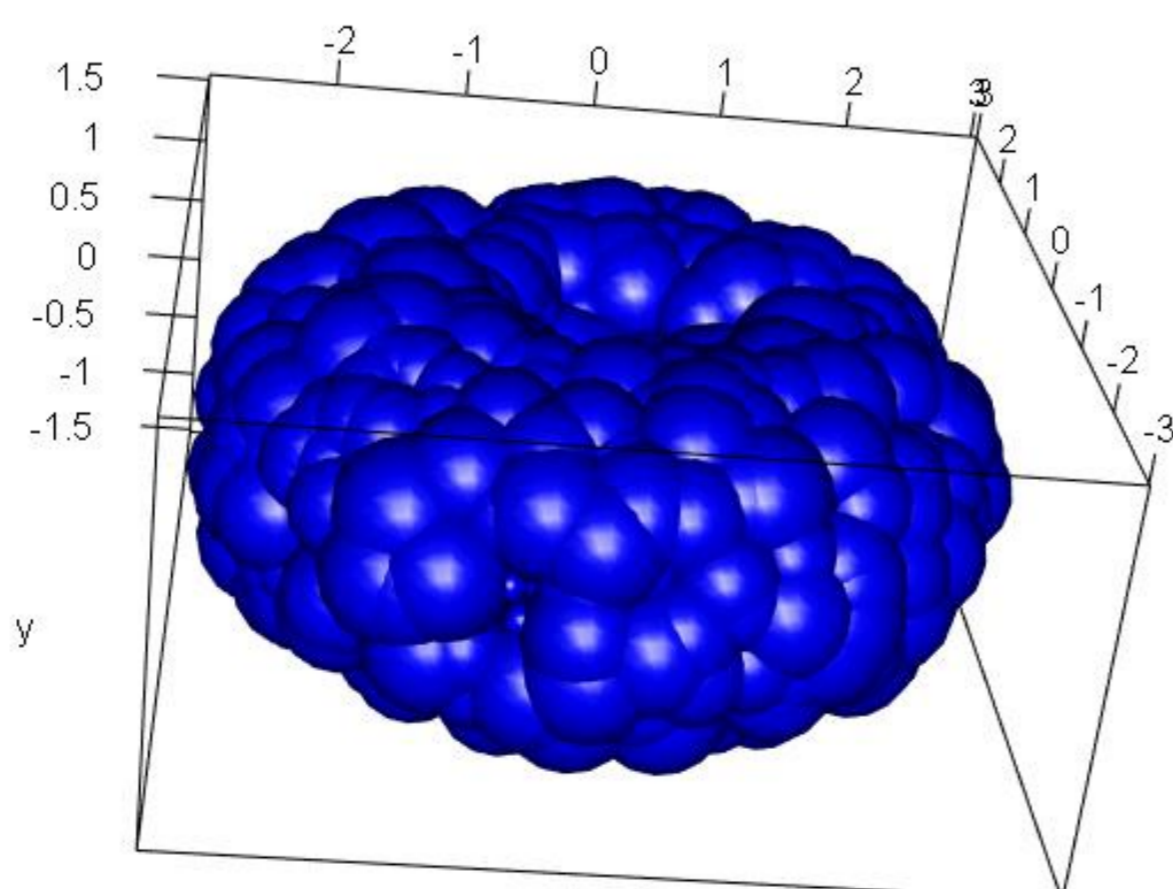
What, exactly, does "uniform" mean?

- Uniform with respect to a measure (sets with same measure have same probability)
- Length, Area, Volume, ...

$$\mathcal{H}^m(A) = \lim_{\delta \rightarrow \infty} \inf_{A \subseteq \cup B_i, \text{diam}(B_i) \leq \delta} \sum \alpha_m \left(\frac{\text{diam}(B_i)}{2} \right)^m$$

Volume of unitary (m-1)-sphere α_m

Countable cover of A



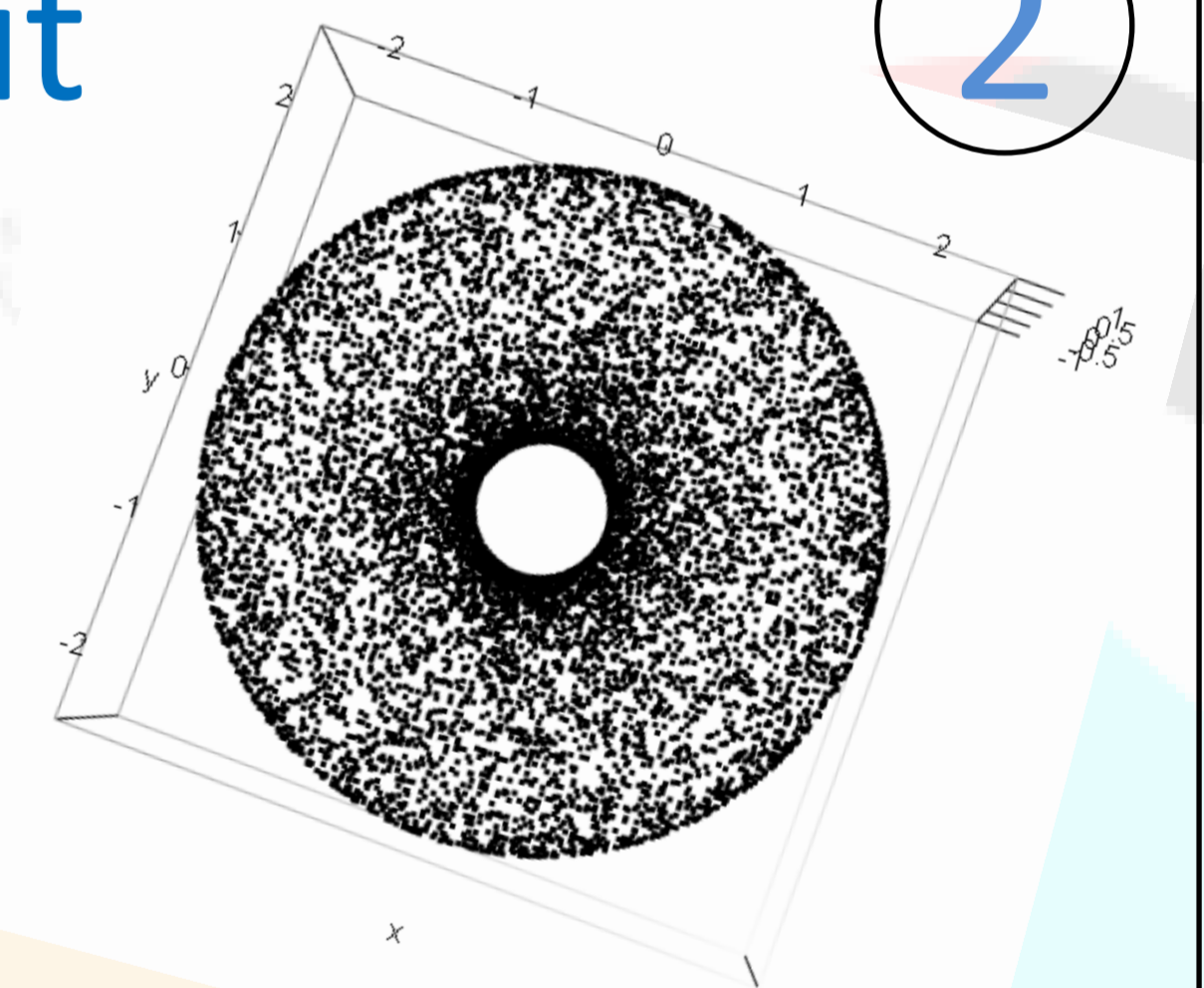
Torus covered by 2-spheres

A simple and natural method

Sprinkling simply, but non-uniformly

2

Sampling uniformly from the domain of the parametrization



- Tempting to use
- Eventually covers the manifold
- Linear complexity on sample size
- If $J_m f$ is constant we are sampling \mathcal{H}^m -uniformly

Torus: Regions with highest concentration have highest Gaussian curvature.

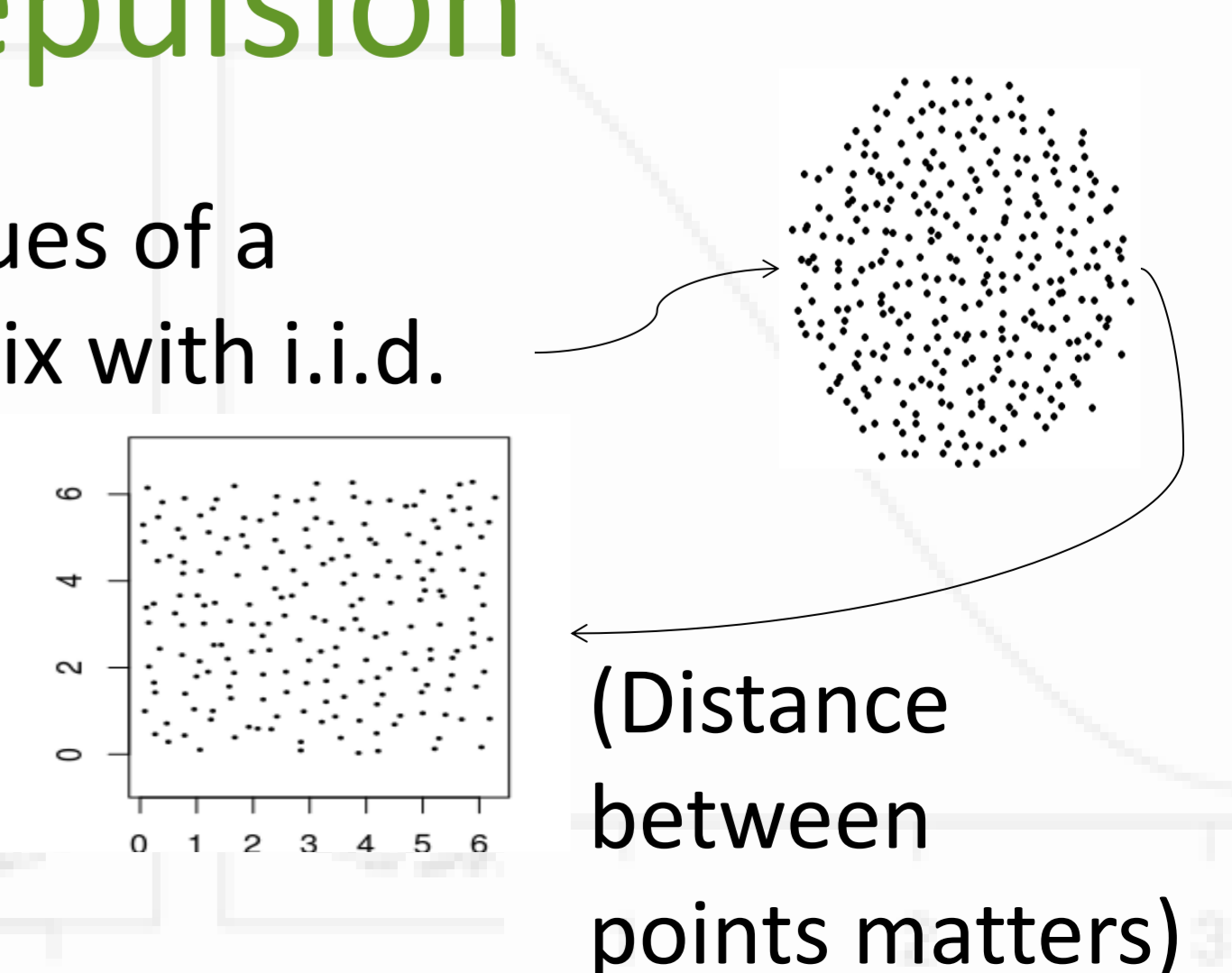
A useful method for TDA

Sprinkling with repulsion

3

Our proposal: Use eigenvalues of a random matrix with i.i.d. entries

- Slow
- Based on Circular Law for random matrices [1]



(Distance between points matters)

Asymptotic behavior

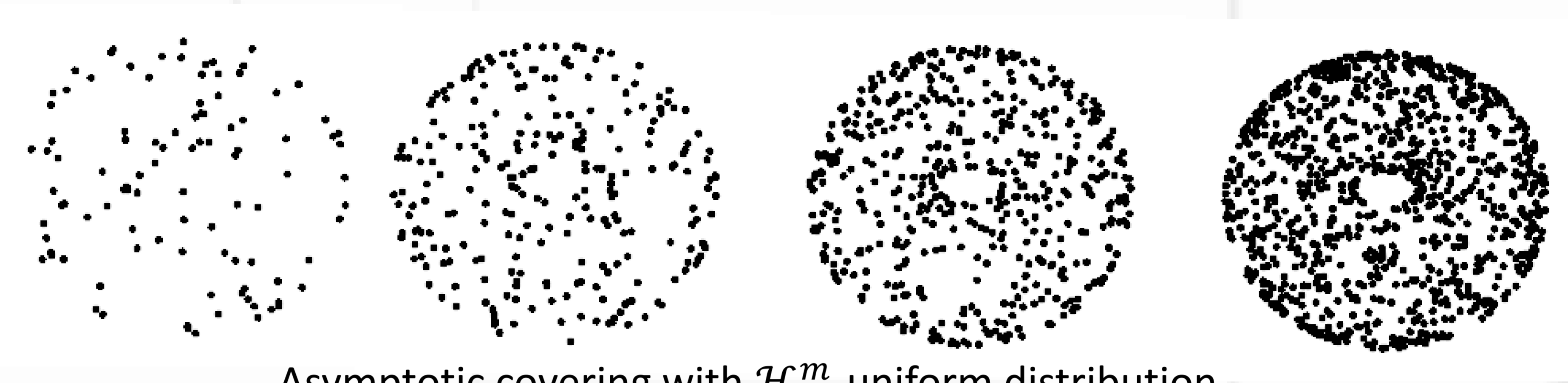
Application of concentration inequality in Fasy et al. [3]

Bottleneck distance d_A , function distance to closed subset A Hausdorff distance

$$\mathbb{P}\left(W_\infty\left(\text{dgm}(d_{S_n}), \text{dgm}(d_M)\right) > t\right) \leq \mathbb{P}\left(H(S_n, M) > t\right) \leq \frac{2^d}{\rho(t/2)t^d} e^{-n\rho(t)t^d}$$

$\rho(x, t) = \frac{\mathbb{P}(B_x(t/2))}{t^d}$
 $\rho(t) = \inf_{x \in M} \rho(x, t)$
 n points from a distribution on M
 Assumptions about how small it can be (bounded away from zero)

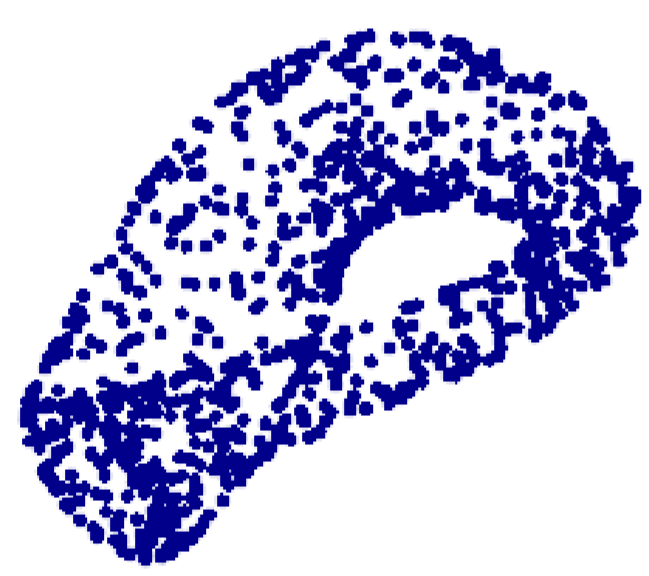
Persistence diagrams Stability theorem



Examples

Klein Bottle

(Using a parametrization for the Klein Bottle in [4])



Uniform on domain

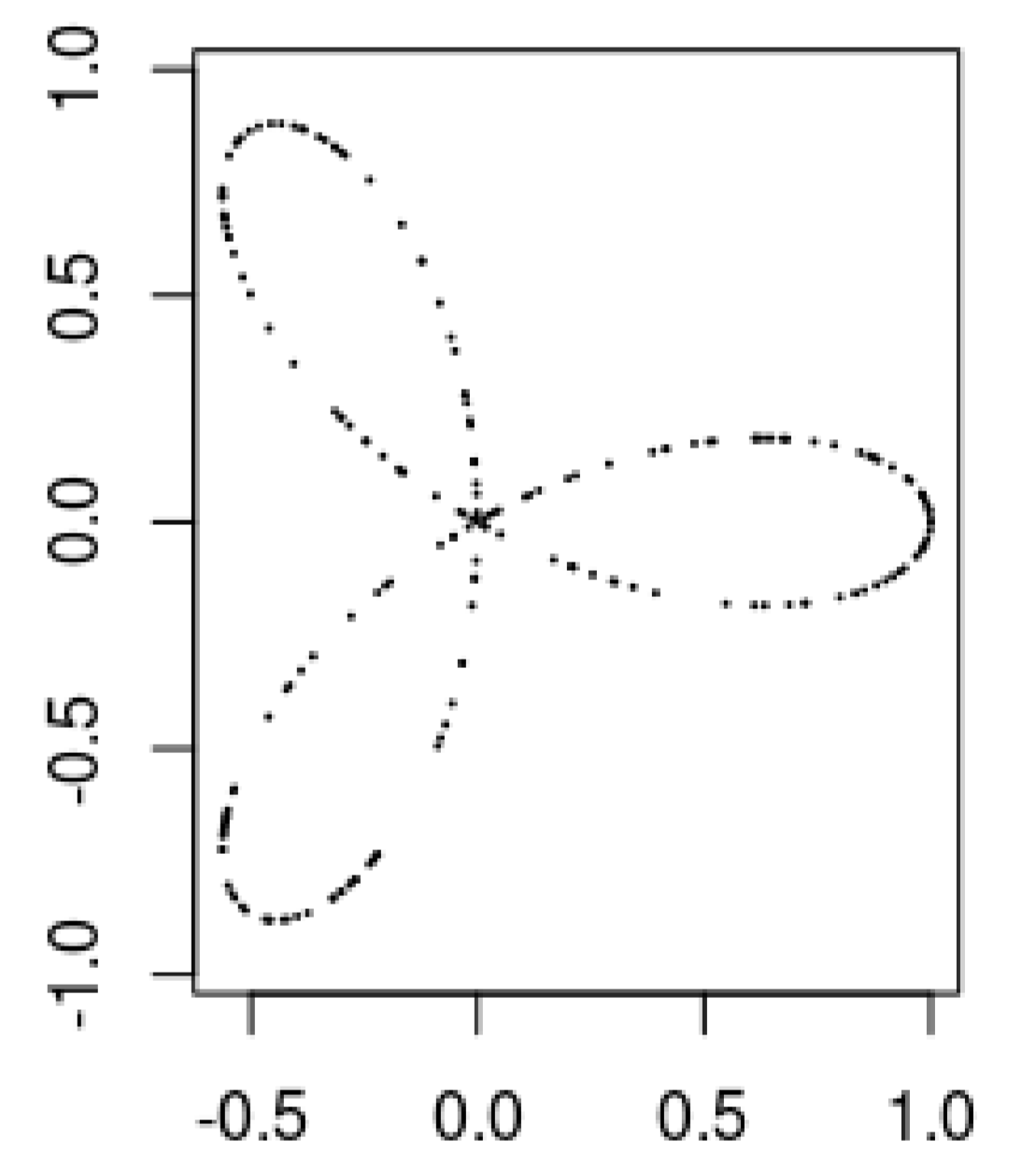


\mathcal{H}^m -uniform



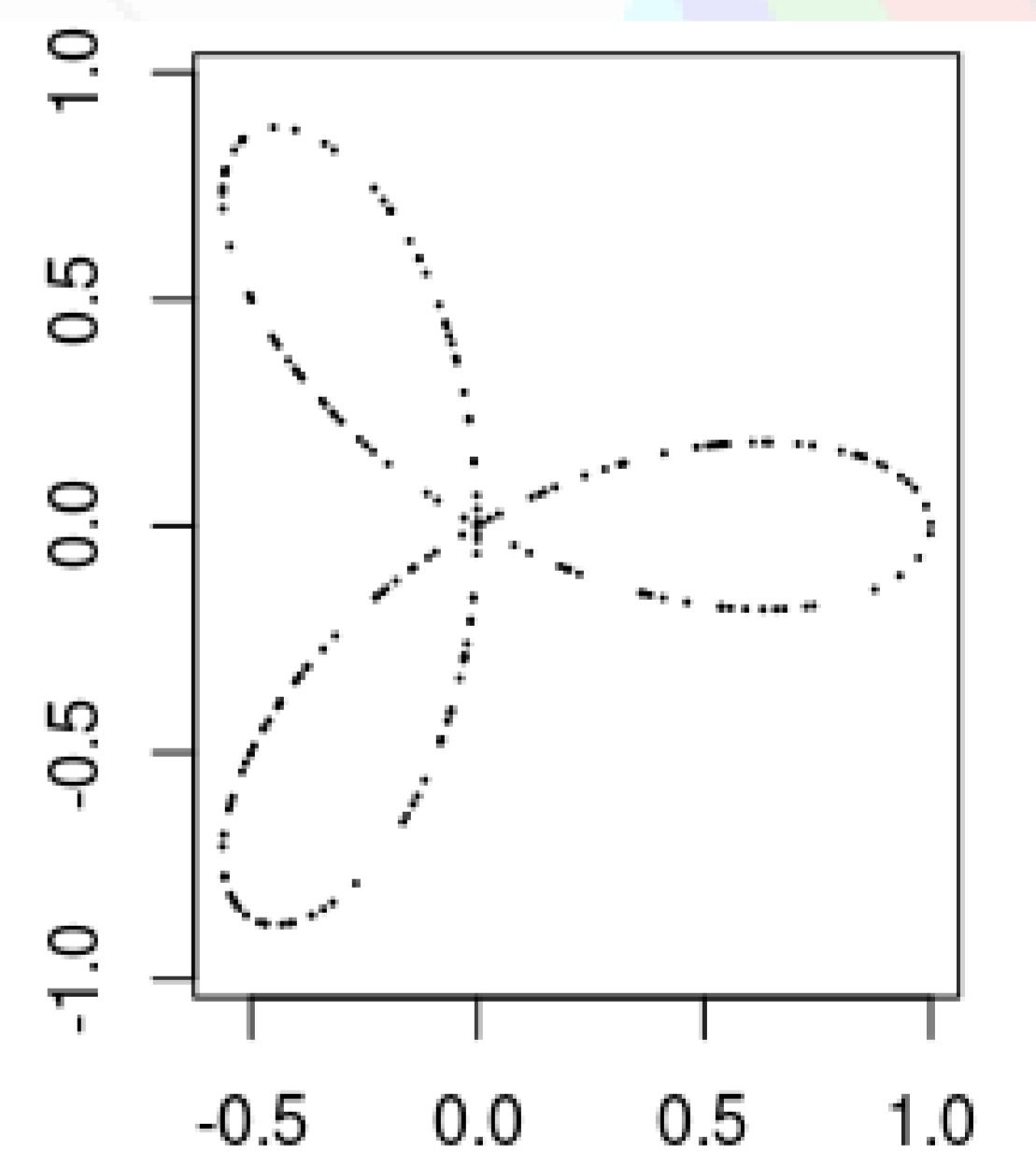
Repulsion

Uniform on domain



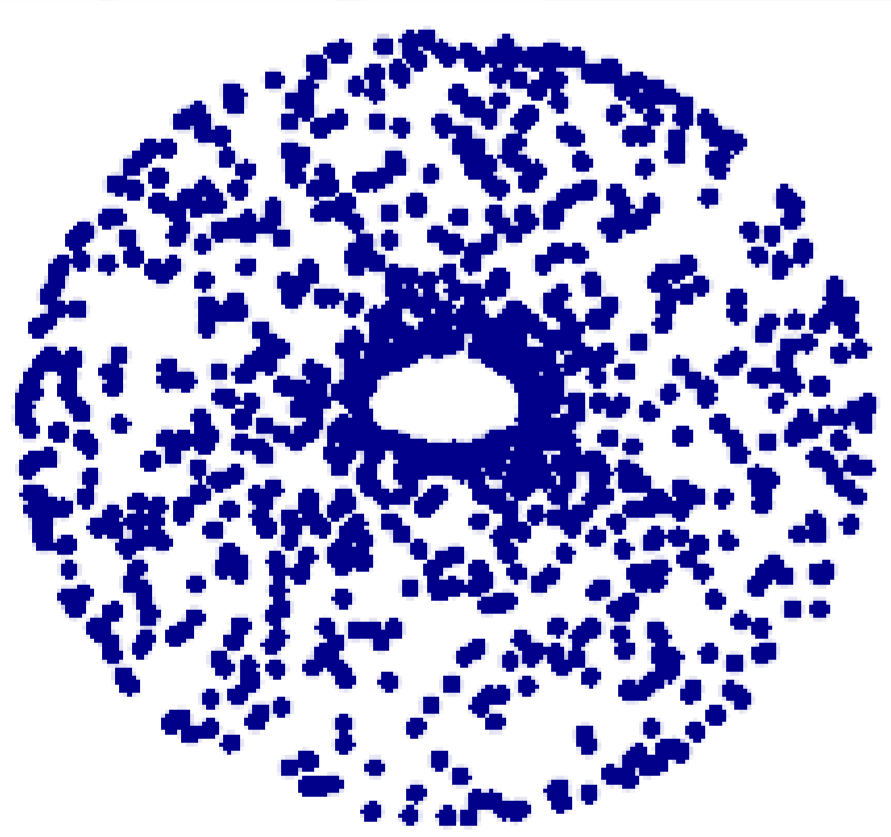
Rose

\mathcal{H}^m -uniform

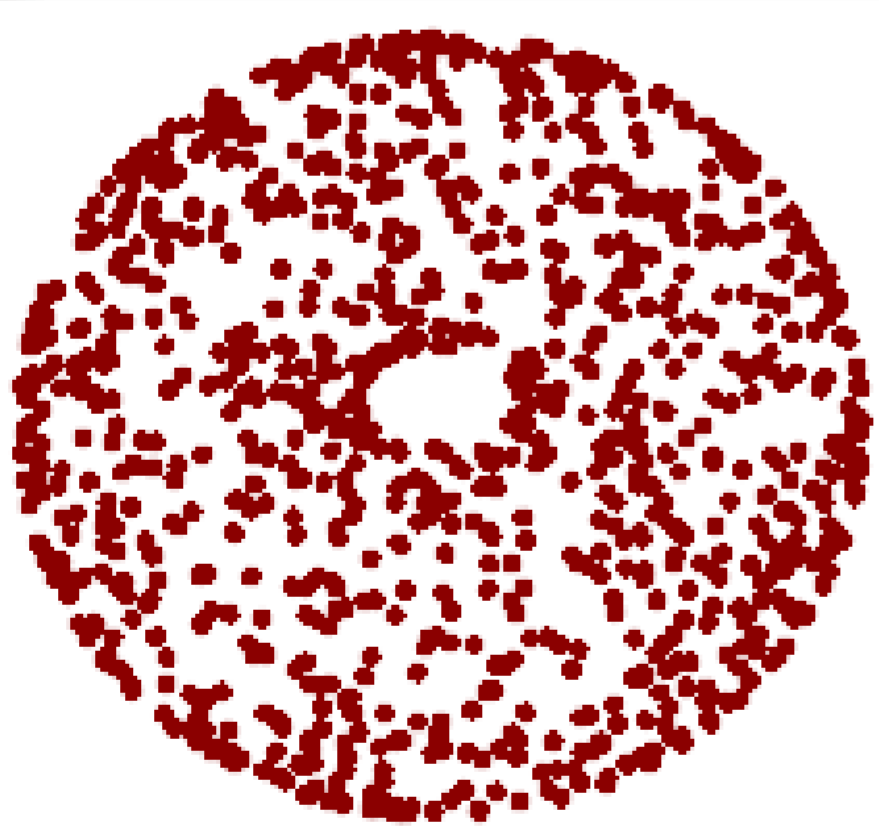


2-Torus

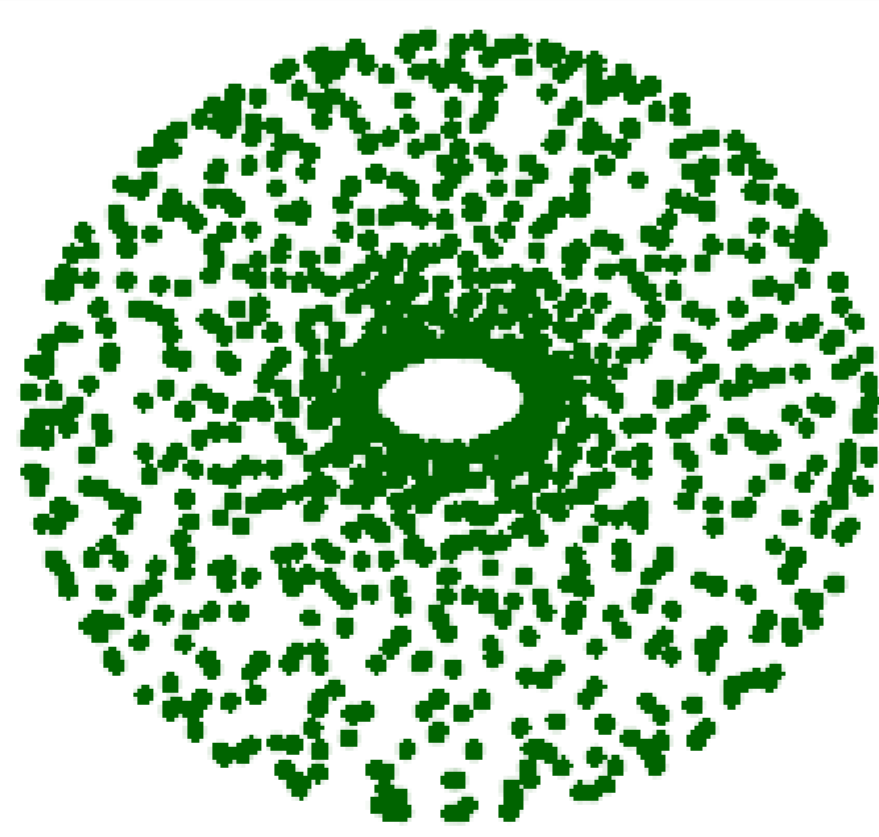
Uniform on domain



\mathcal{H}^m -uniform



Repulsion



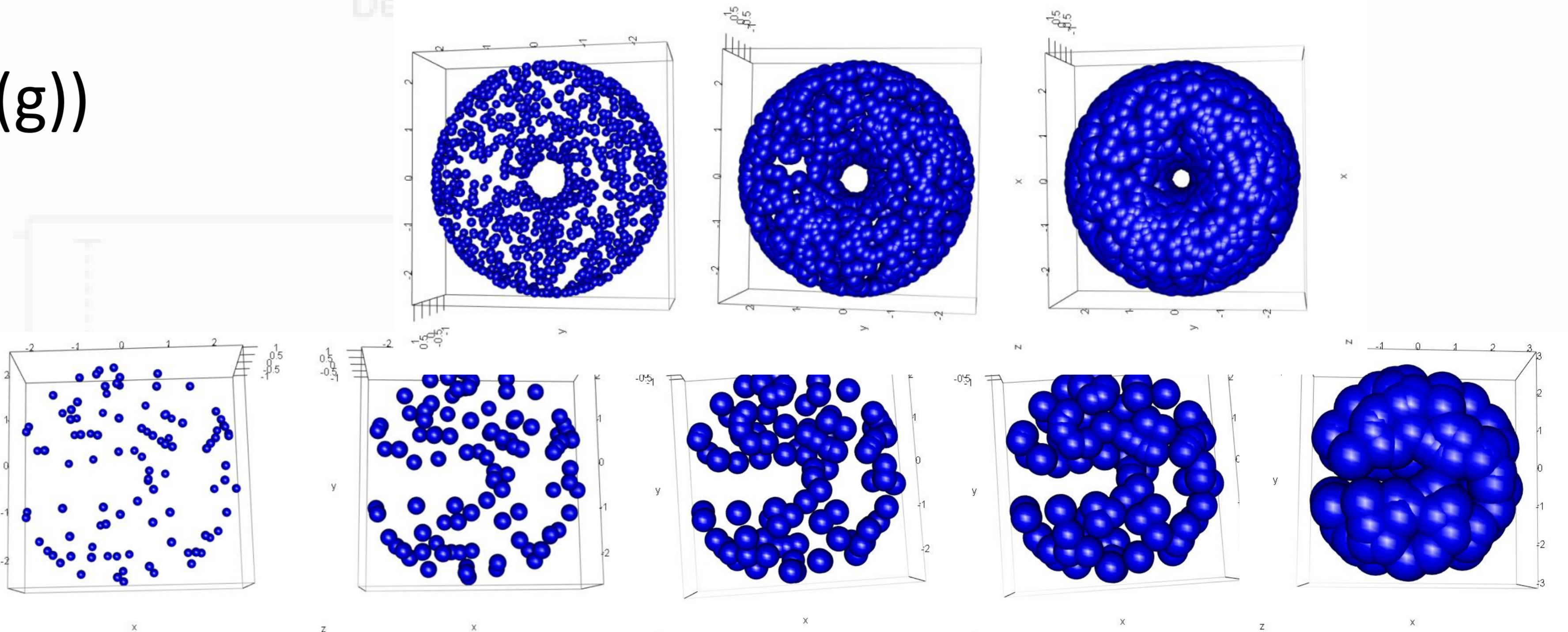
Simulation of PCD* on parametrized manifolds

Topological Data Analysis

Some TDA techniques

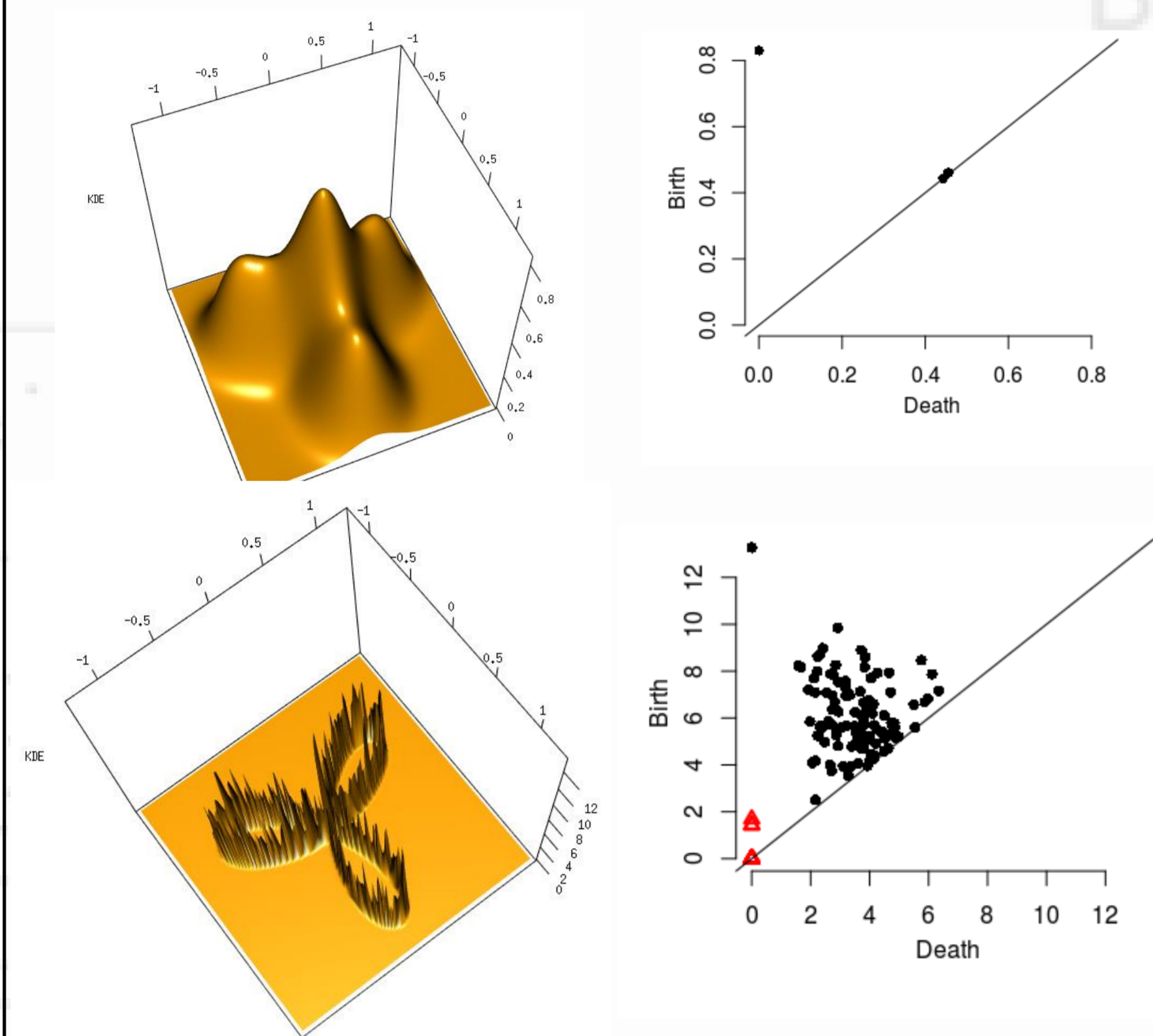
- Real function defined on a manifold or simplicial complex (g)
- Computation of topological features in each upper level or sublevel sets (Betti numbers)
- Summarizes critical points ($dgm(g)$)

Distances (Vietoris-Rips filtration)

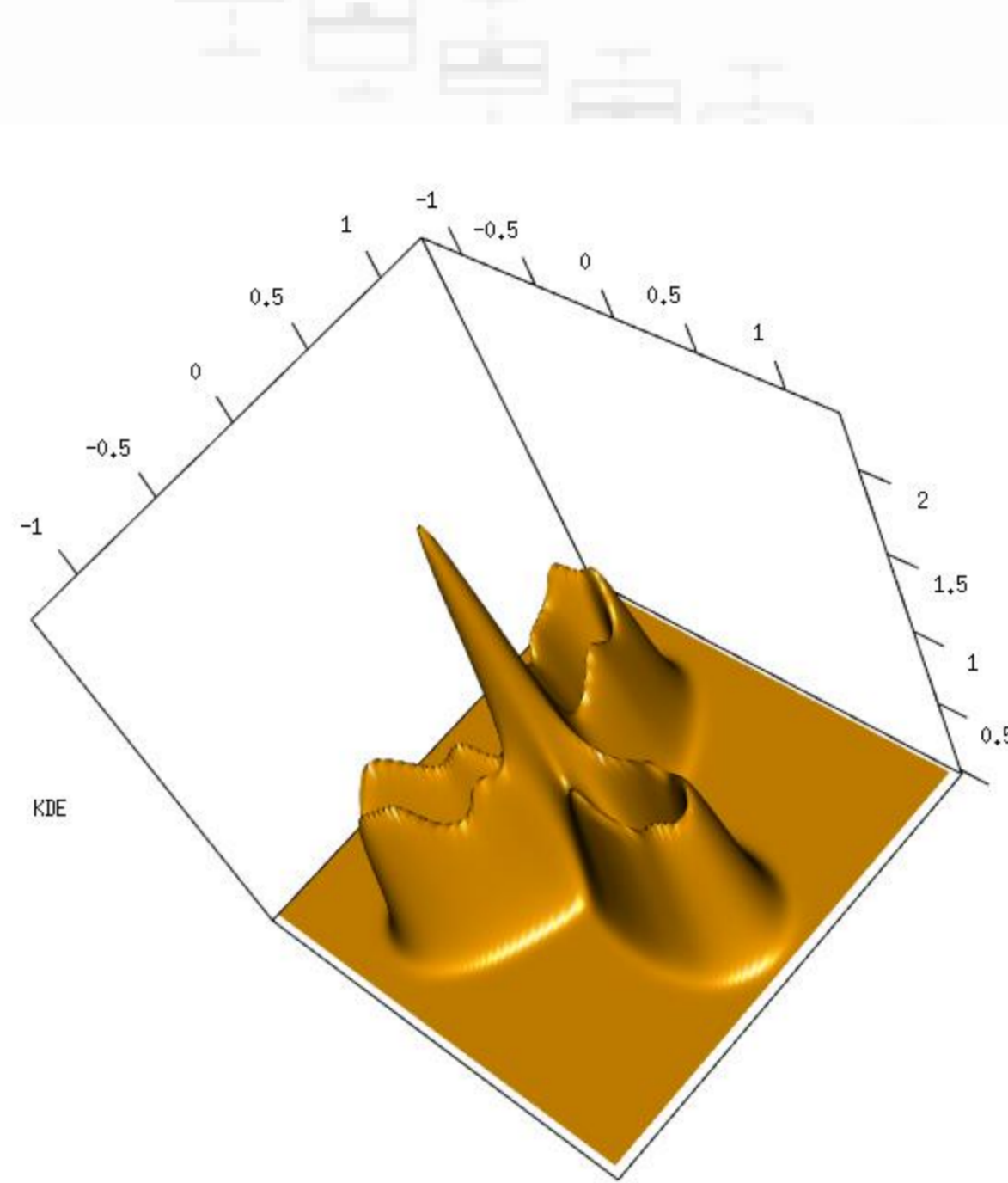


Kernel density estimator (KDE)

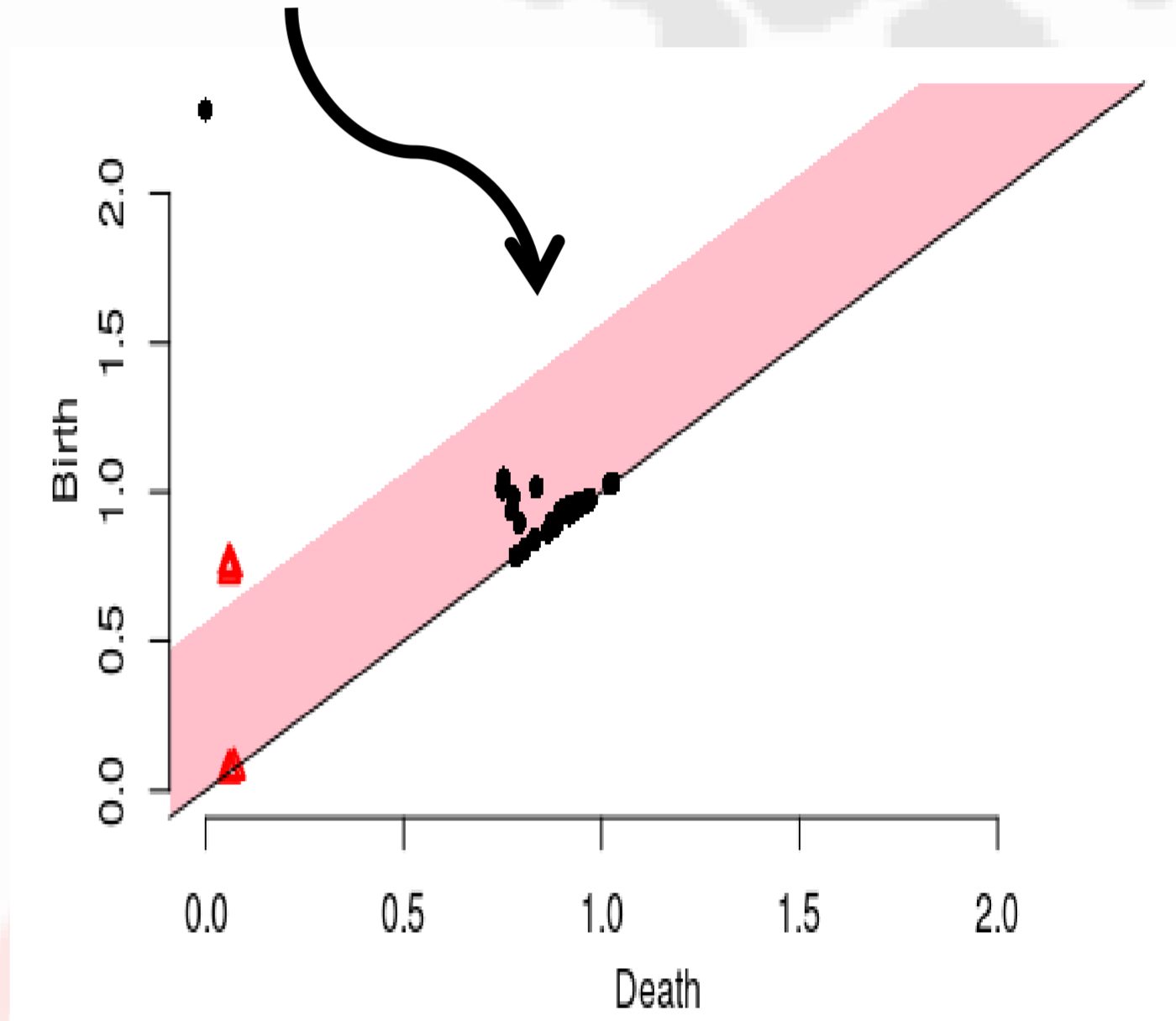
Choose carefully smoothing parameter



An Illustrative example

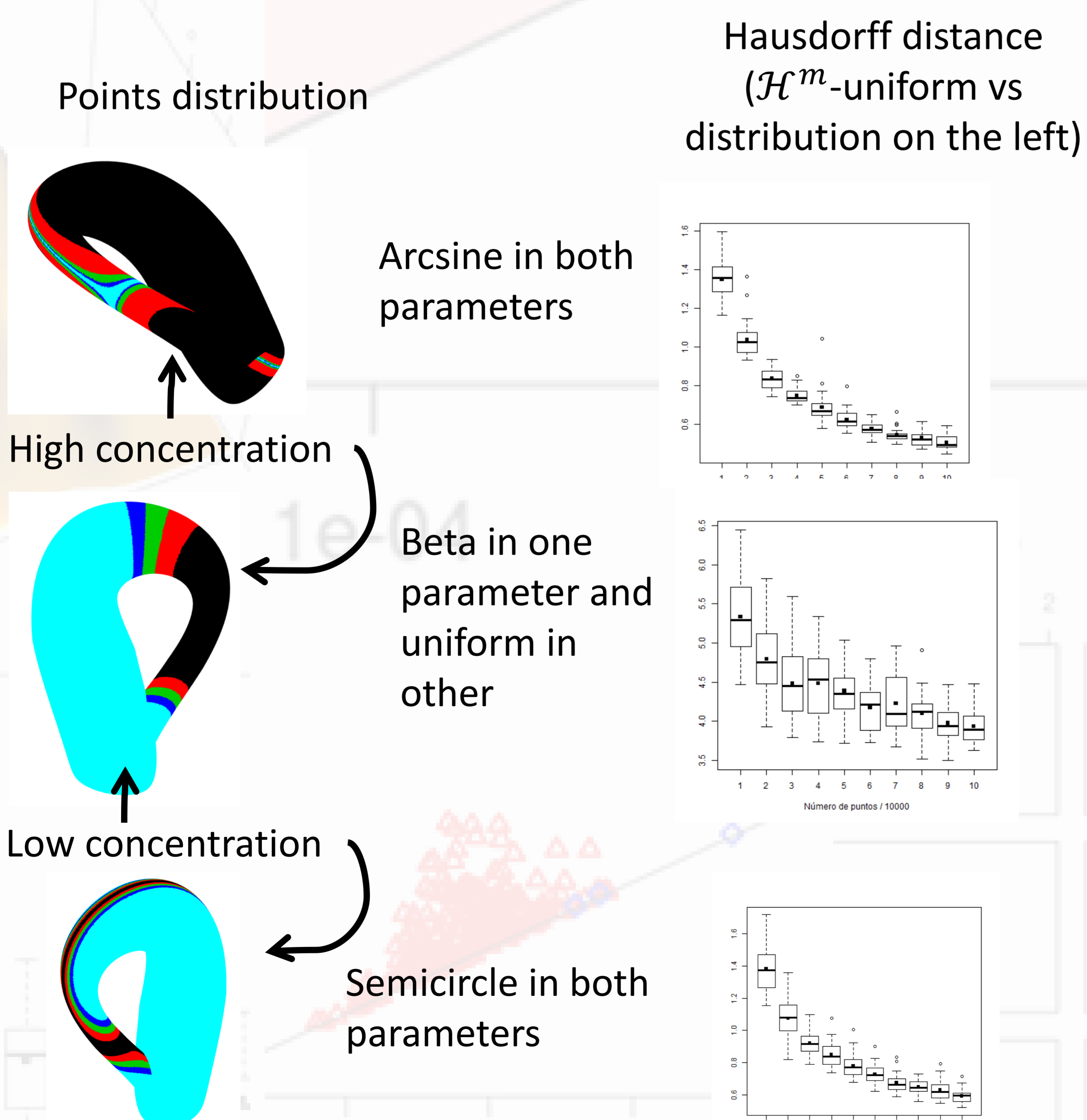


Confidence band

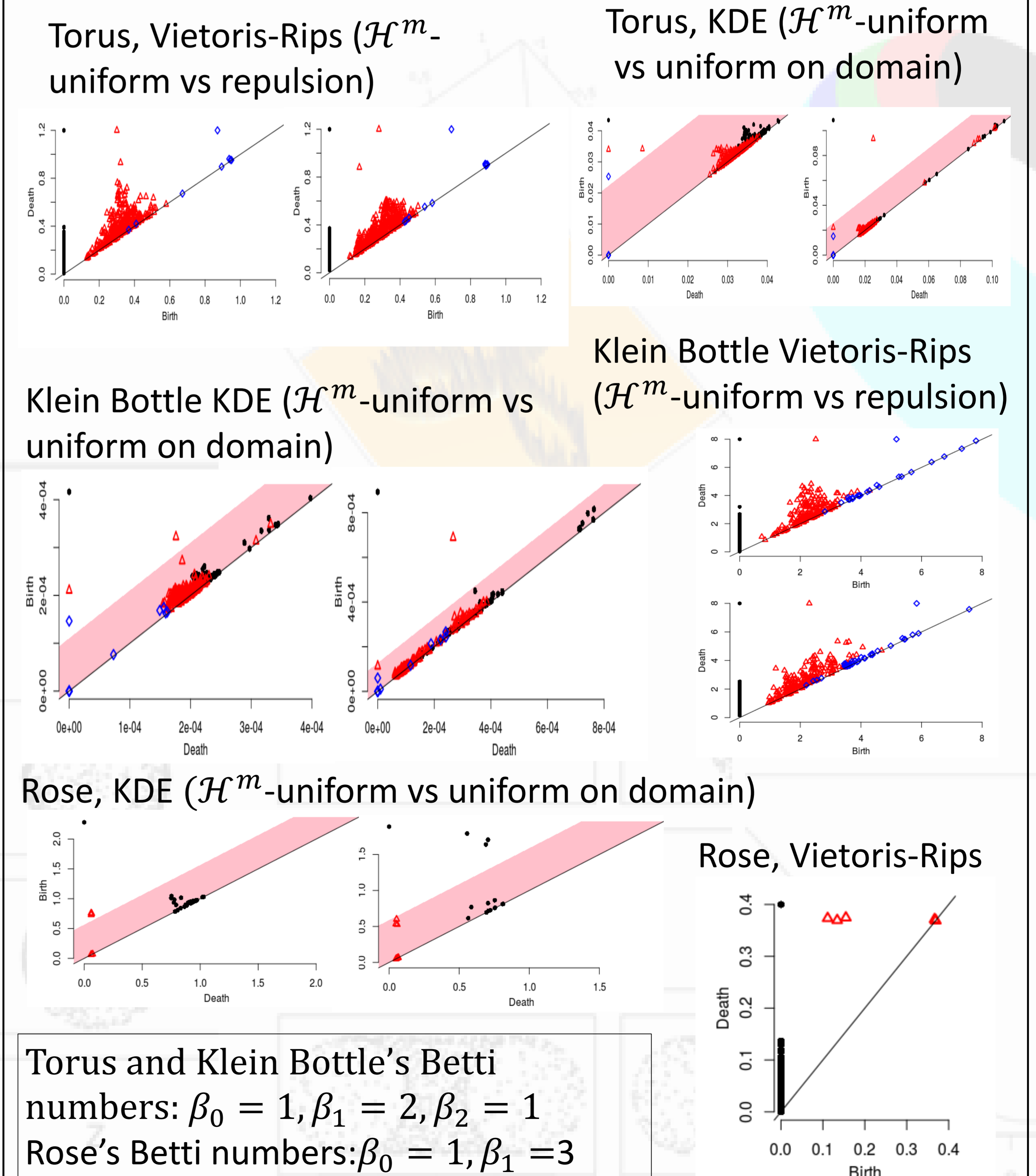


Examples on a Klein Bottle

Sampling from other distributions on domain.



TDA applied to simulated PCD



Conclusions and forthcoming work

- Uniform on the domain is NOT equivalent to uniform on the manifold (exception: constant Jacobian).
- Choose method depending on objectives and other factors like computing cost.
- For uniform distribution on the manifold use parametrizations with a “simple” Jacobian.
- For examples in torus, simulating with repulsion yields better persistence diagrams.
- When using filtration over KDE: best result obtained with uniform distribution on manifold.
- Sampling from other distributions gives empirical insight in the identification of regions with highest and lowest probability, and on the rate of convergence in the concentration inequality since the computation of $\rho(t)$ is not in general easy.

References

1. Bordenave, C. and D. Chafaï: Lecture notes on the circular law. Proceedings of Symposium in Applied Mathematics, 72:1–34, January 2014.
<http://dx.doi.org/10.1090/psapm/072/00617>.
2. Diaconis, P., S. Holmes, and M. Shahshahani: *Sampling from a Manifold*. Advances in Modern Statistical Theory and Applications: a Festschrift in honor of Morris L. Eaton, 10:102–125, 2013.
3. Fasy, B.T., F. Lecci, A. Rinaldo, L. Wasserman, S. Balakrishnan, and A. Singh: *Confidence Sets for Persistence Diagrams*. The Annals of Statistics, 42(6):2301–2339, 2014.
4. Franzoni, G. *The Klein bottle: Variations on a theme*. Notices of the AMS, 59(8):1094–1099, 2012.

Acknowledgments

To Victor Pérez-Abreu for his constant support and guidance; Carlos Vargas for his suggestion for considering repulsion; and to Miguel Nakamura for useful comments.

Contact info

yair.hernandez@cimat.mx
gilberto.flores@cimat.mx

Partially supported by
Conacyt-SNI 4337.