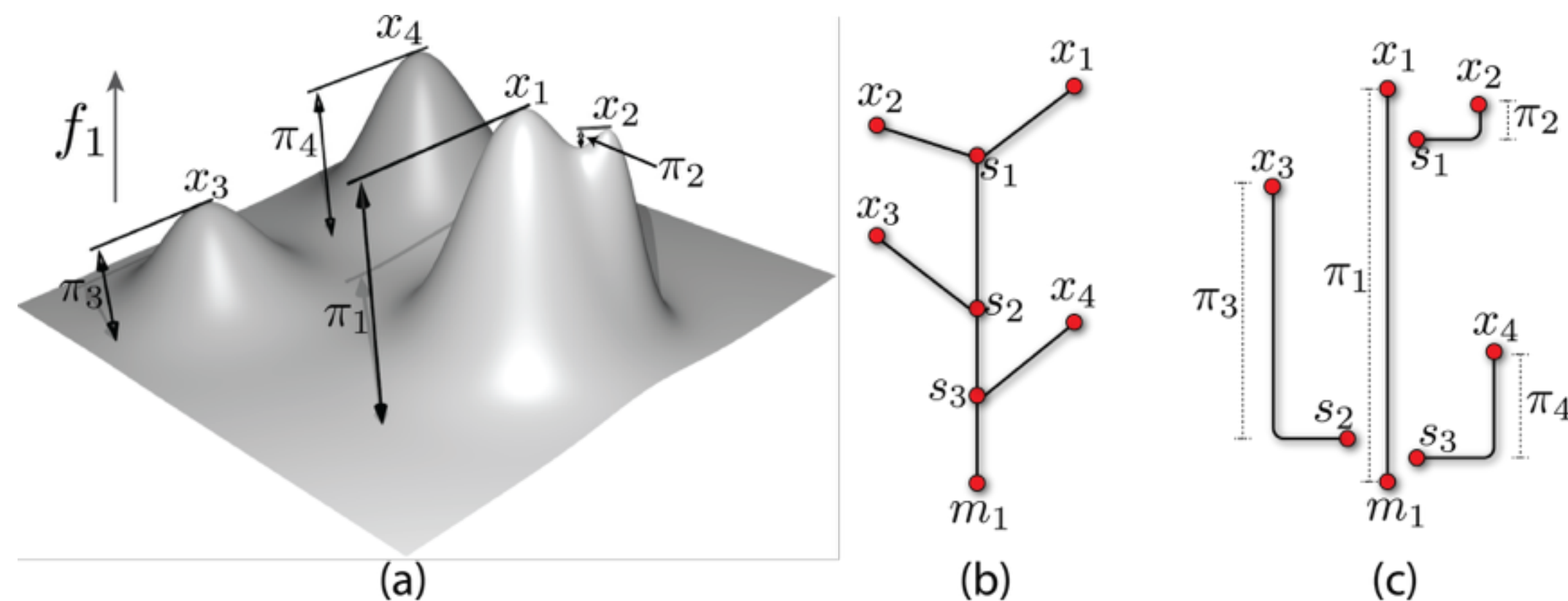


# Using Maximum Topology Matching to Explore Differences in Species Distribution Models



**Jorge Poco**  
Harish Doraiswamy  
Marian Talbert  
Jeffrey Morisette  
Cláudio Silva

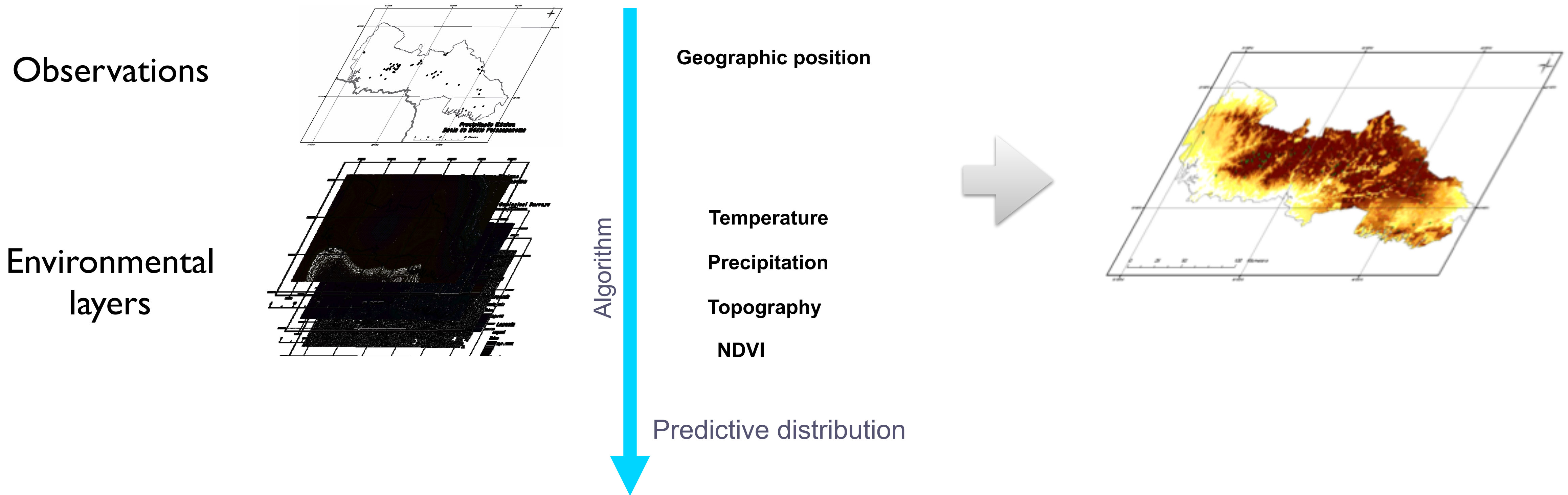


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# Species Distribution Models (SDM)

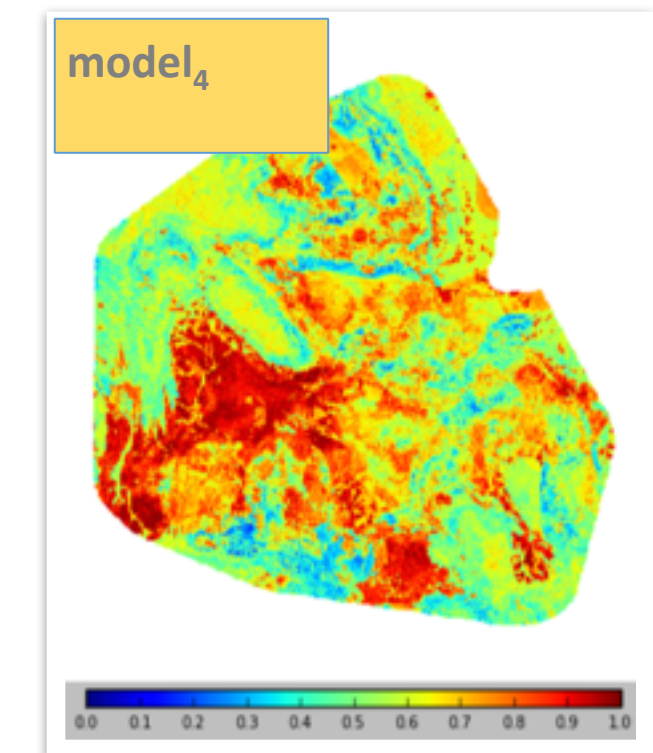
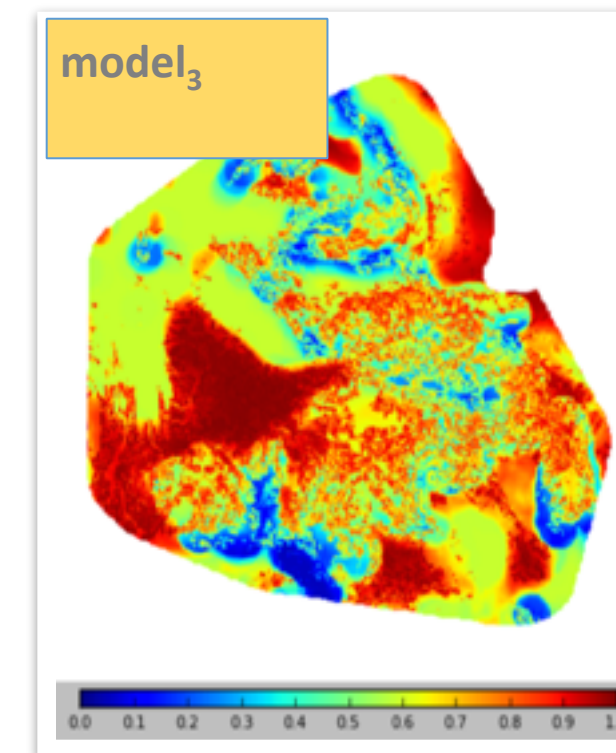
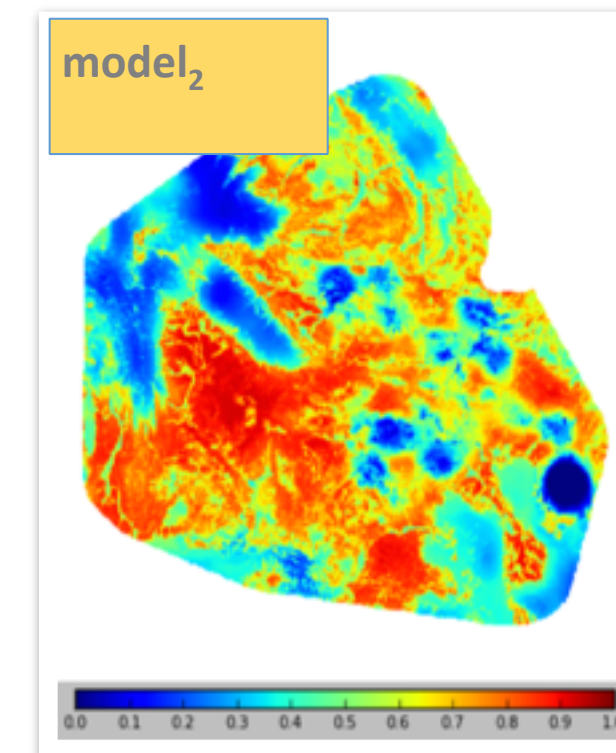
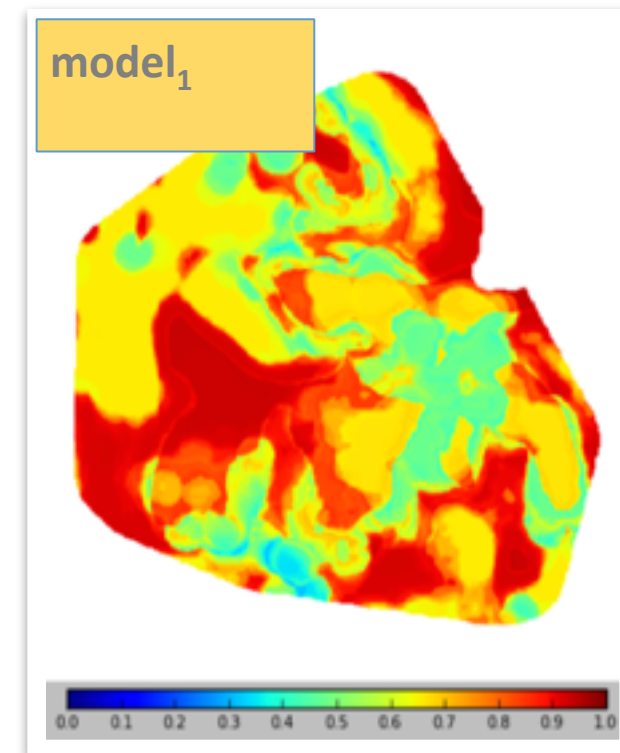


$$f : \mathbb{R}^d \rightarrow [0, 1]$$

# Model Inter-comparison



→  
Predict  
Distribution



Why are they different?  
How do we compare them?

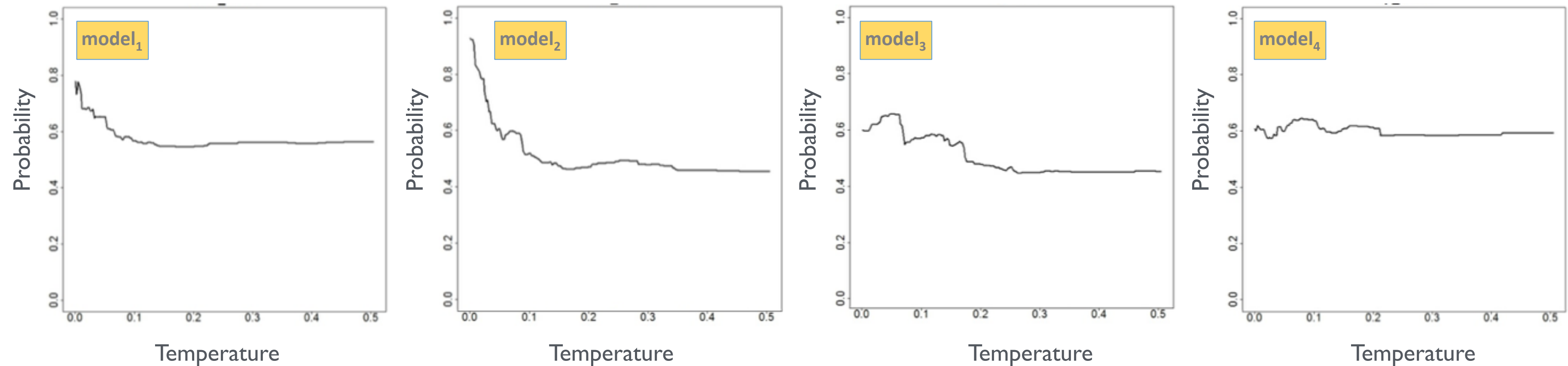


# Ecologist's Approach

Visualize one-dimension at a time (default response curve)

## Predictors

Temperature	[0, 0.5]
Precipitation	mean
Topography	mean
NDVI	mean
...	...



## SHORTCOMINGS:

- Restricts the analysis to one dimension at time
- Interaction between dimensions are lost because of dimensionality reduction



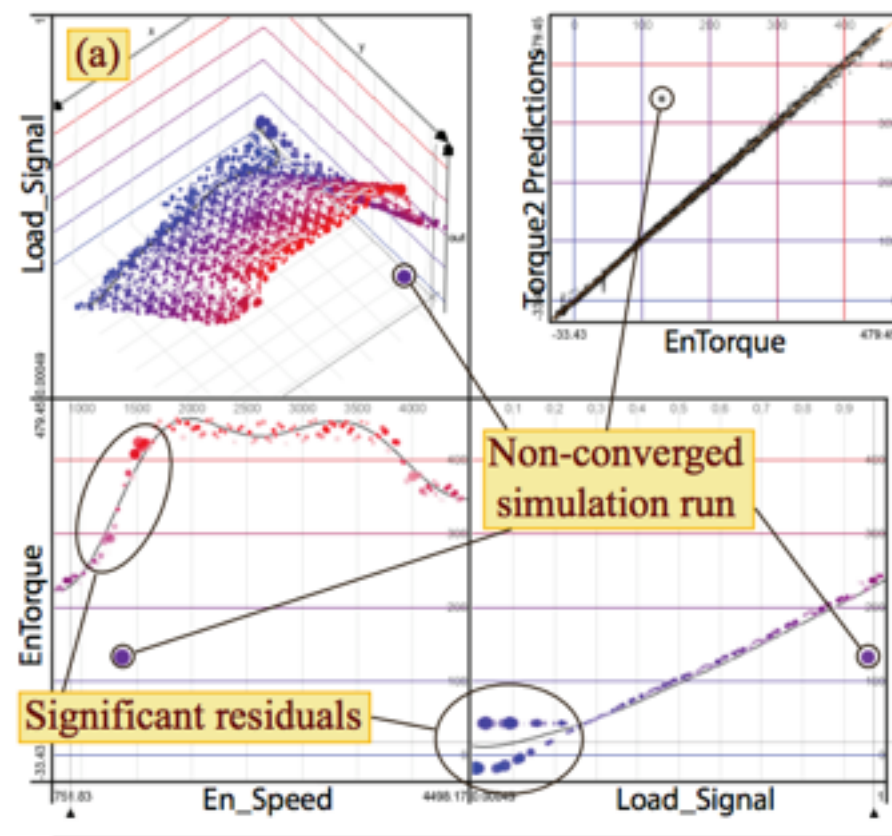
# Goals

- **Explore** SDMs in the high dimensional domain
- **Compare** different SDMs

Our approach is to use the **topology of SDMs** for exploring and comparing them

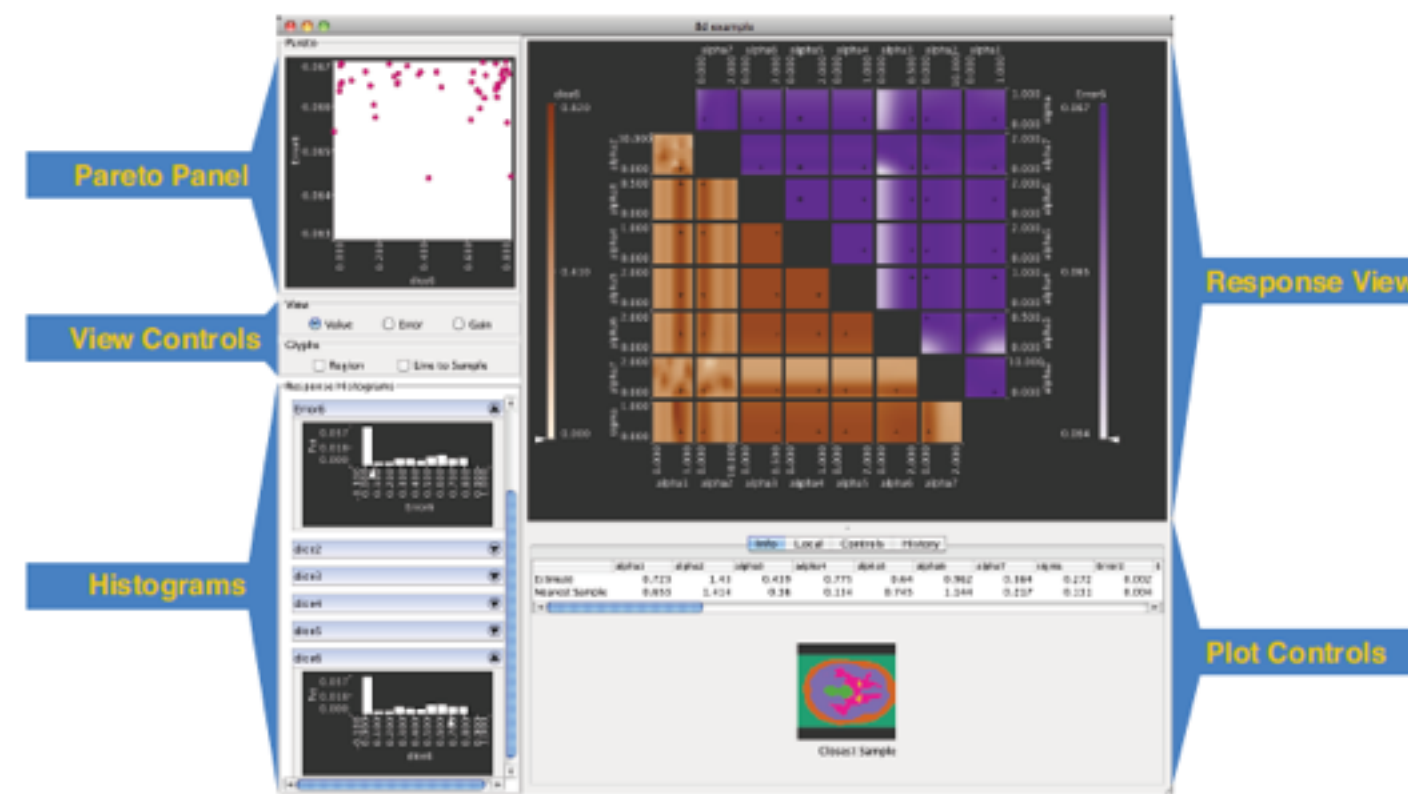
# Challenges: Exploring High Dimensional Functions

[HyperMoVal]



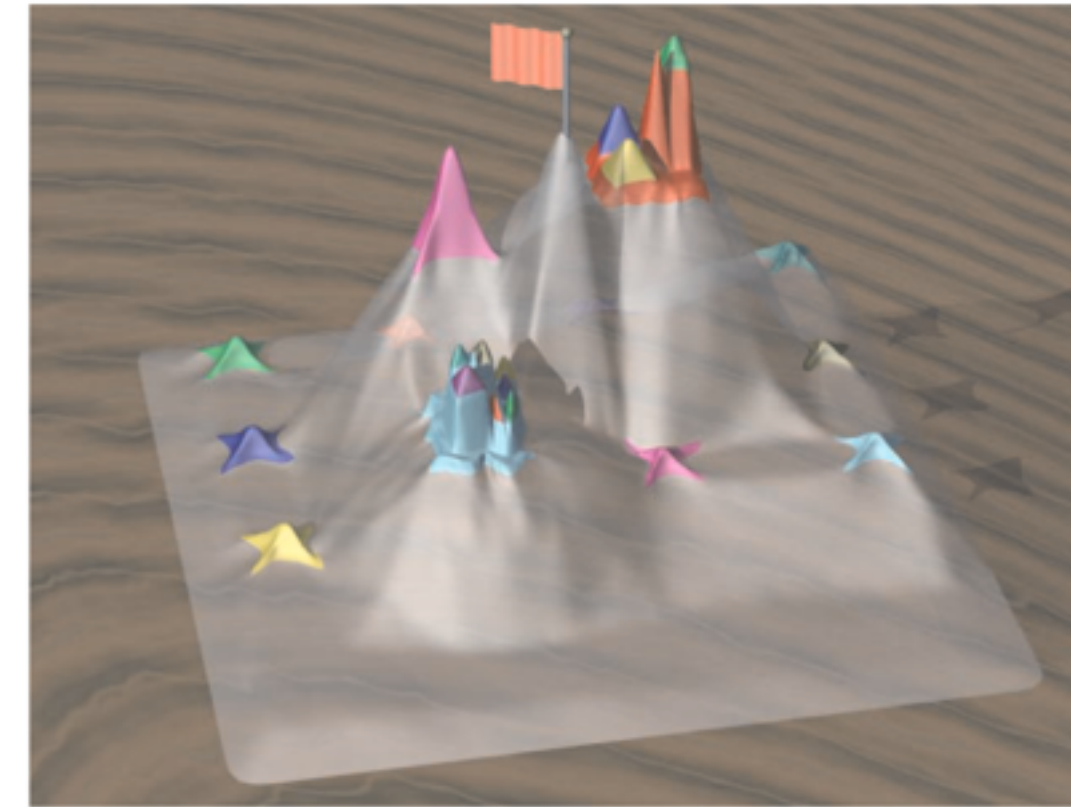
[Piringer et al. 2010]

[Tuner]



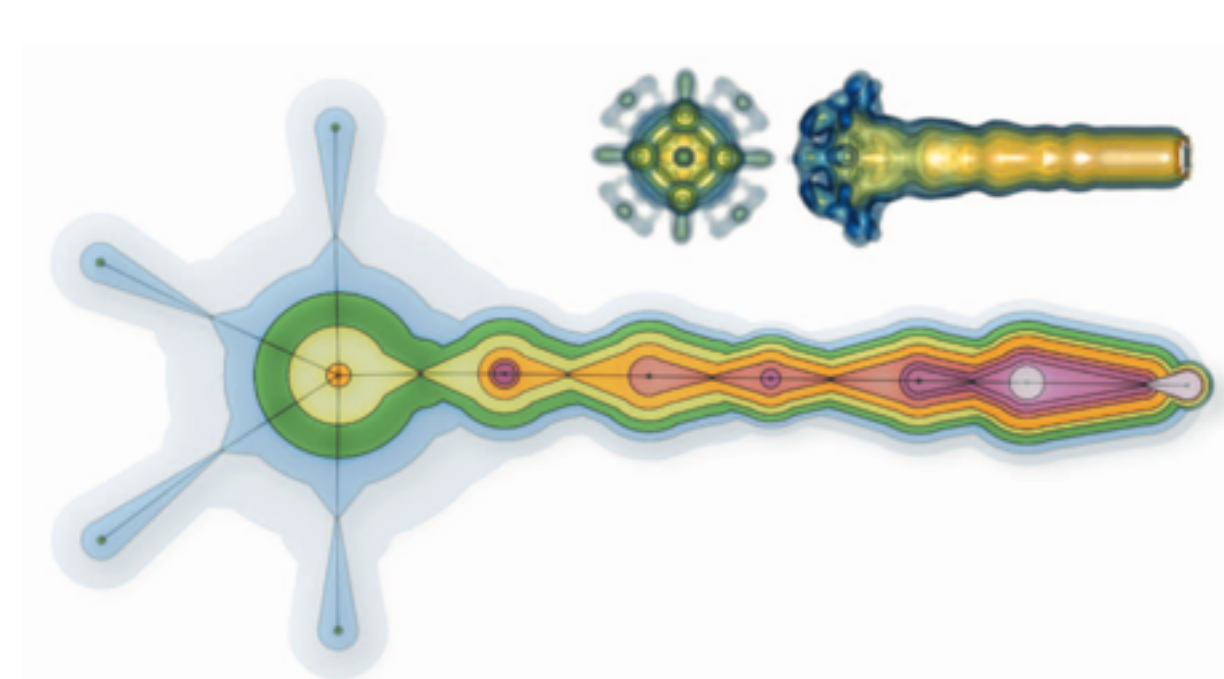
[Torsney-Weir et al. 2011]

[Topological Landscape]



[Weber et al. 2007]

[Topological Spines]



[Correa et al. 2011]

- Shortcoming:
  - Require to manually explore and compare two high dimensional spaces

# Challenges: Comparing Scalar Functions

- Existing work:

- Distance between visual representations of topological persistence

[Carlsson et al. 2004][Cohen-Steiner et al. 2007]



[Carlsson et al. 2004]

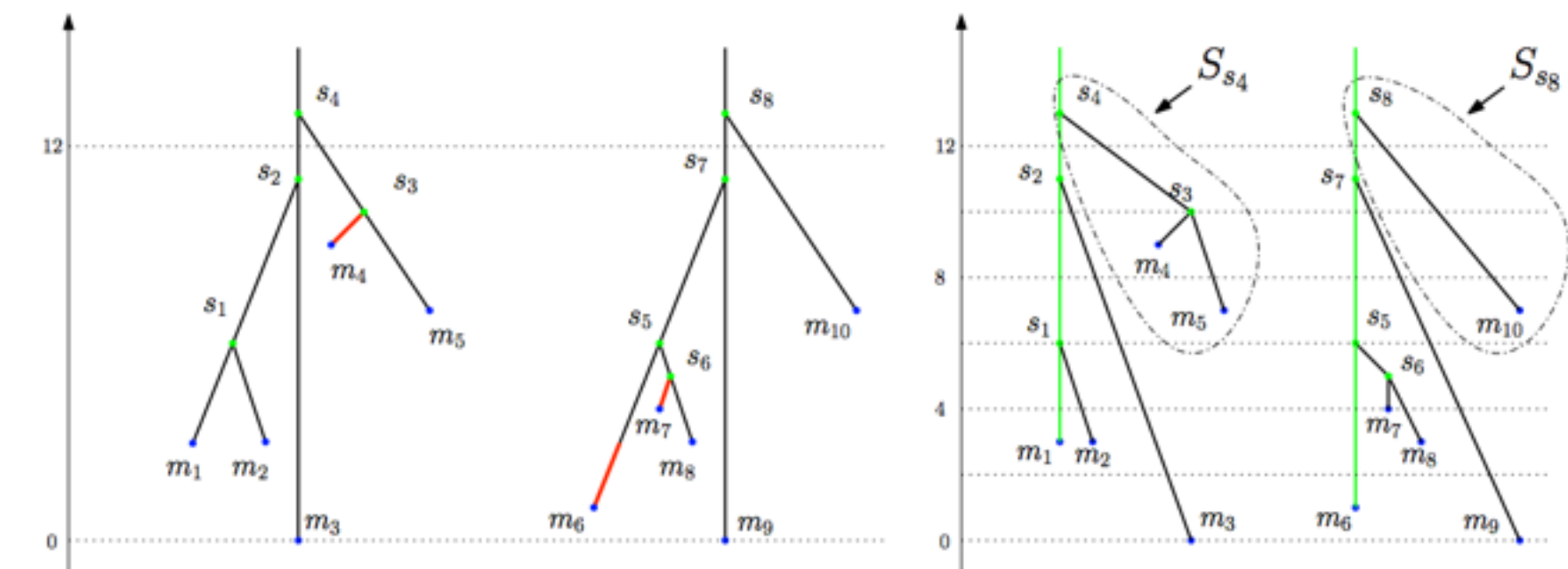
- Distance between topological data structures

- ▶ Merge trees / Reeb graphs

[Morozov et al. 2013][Beketayev et al. 2014][Bauer et al. 2014]

- ▶ Extremum graphs

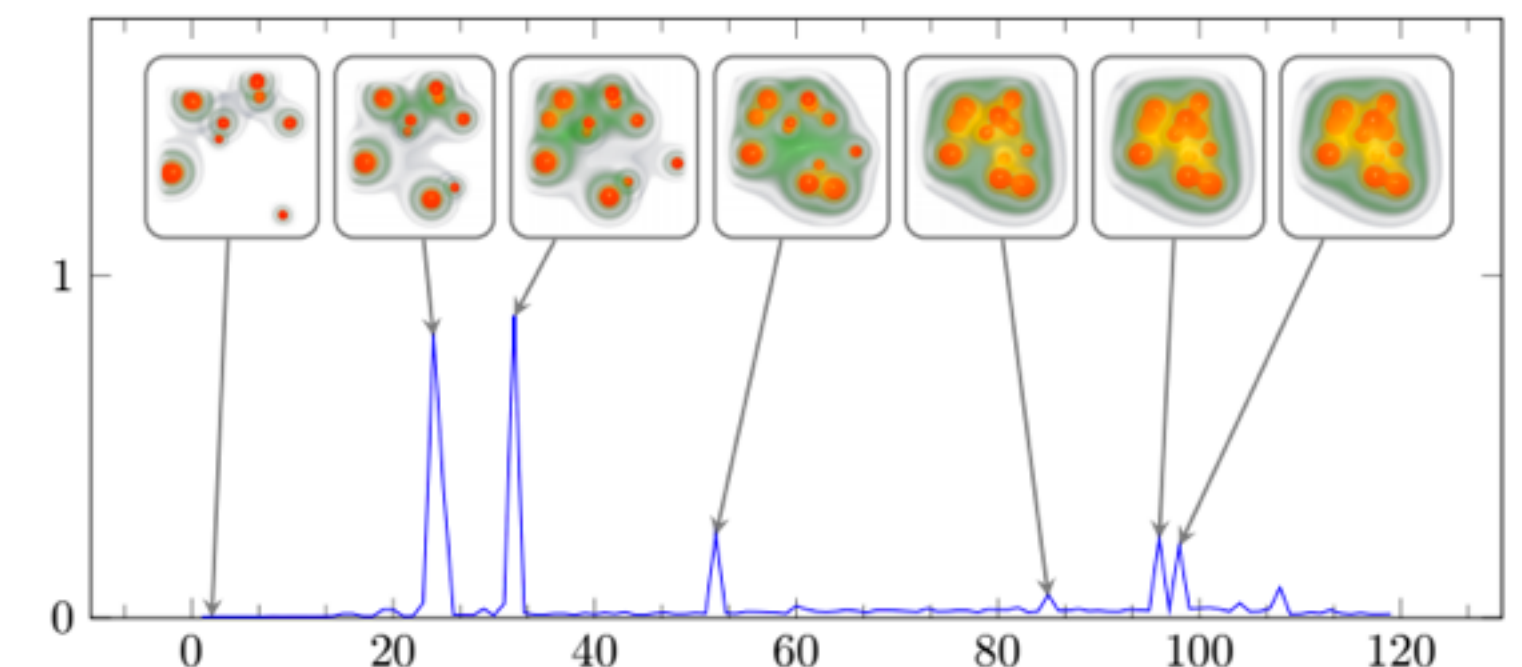
[Narayanan et al. 2015]



[Beketayev et al. 2014]

- Shortcomings:

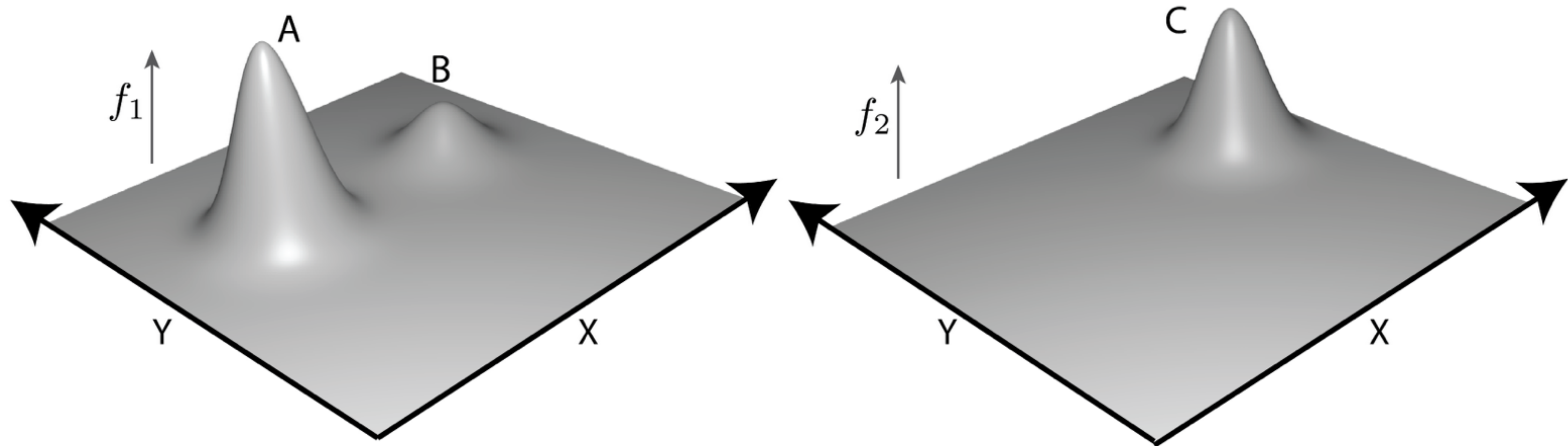
- Location of critical points not considered
- Similar features can be far away



[Narayanan et al. 2015]

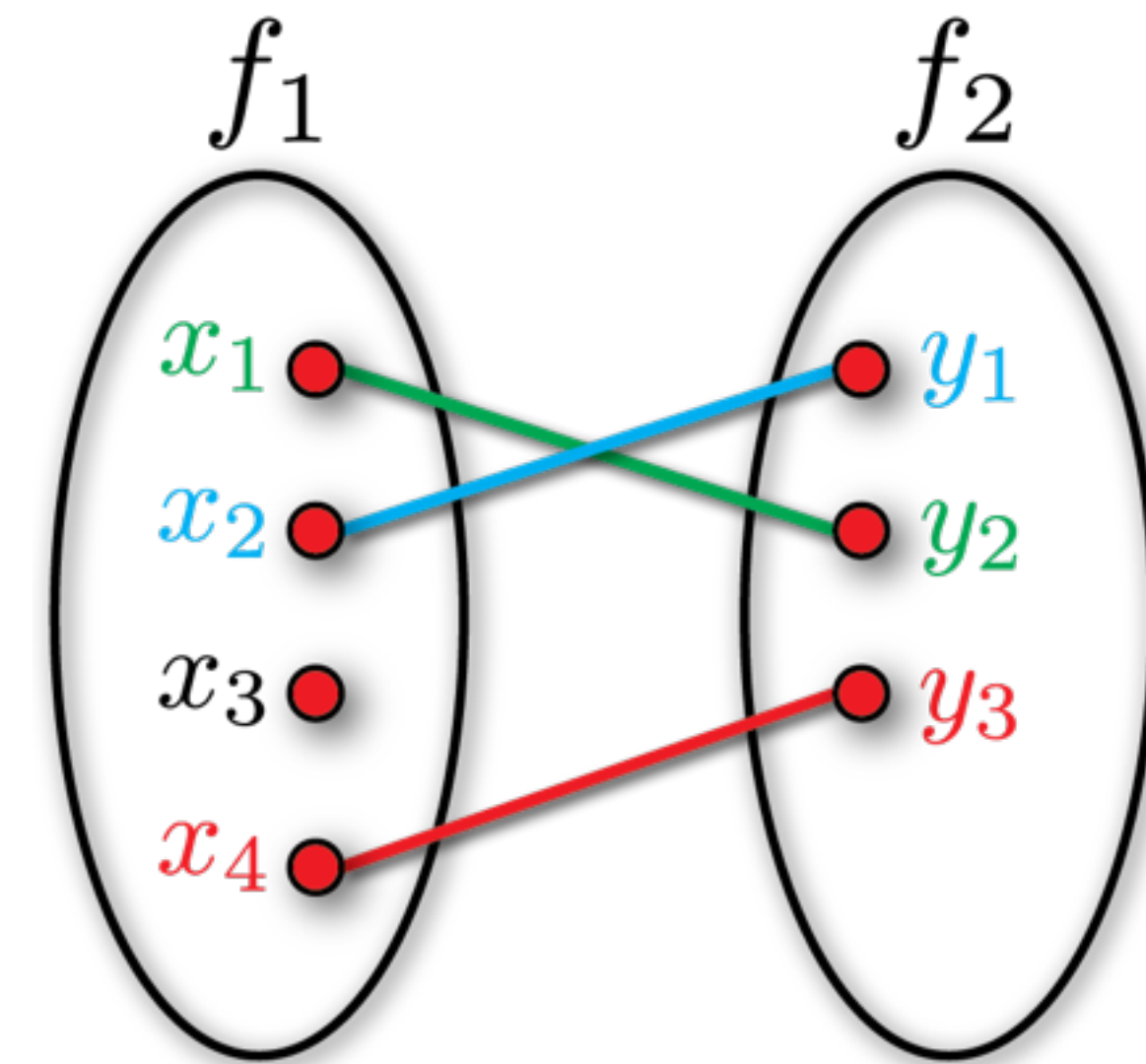
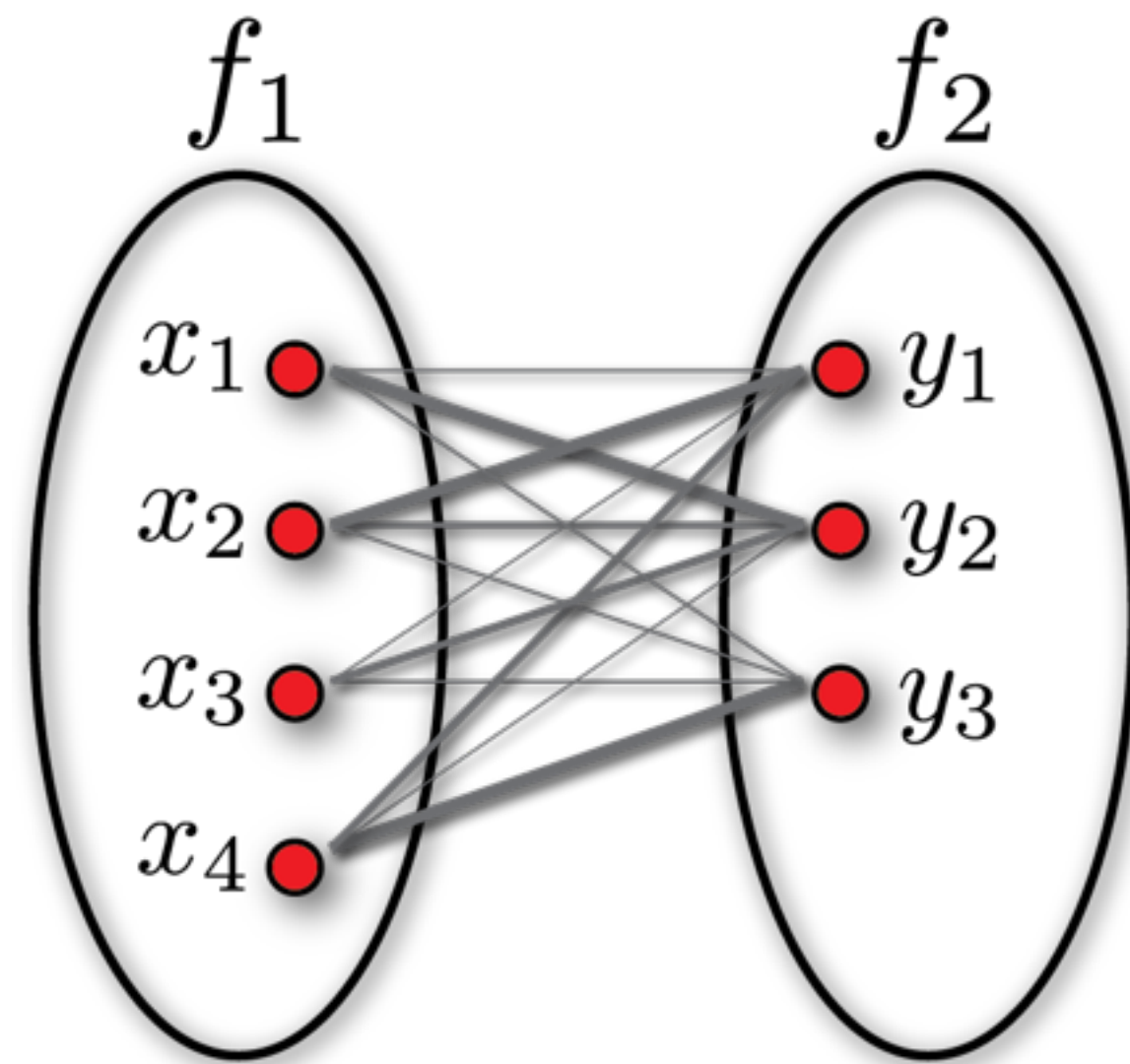


# Motivation



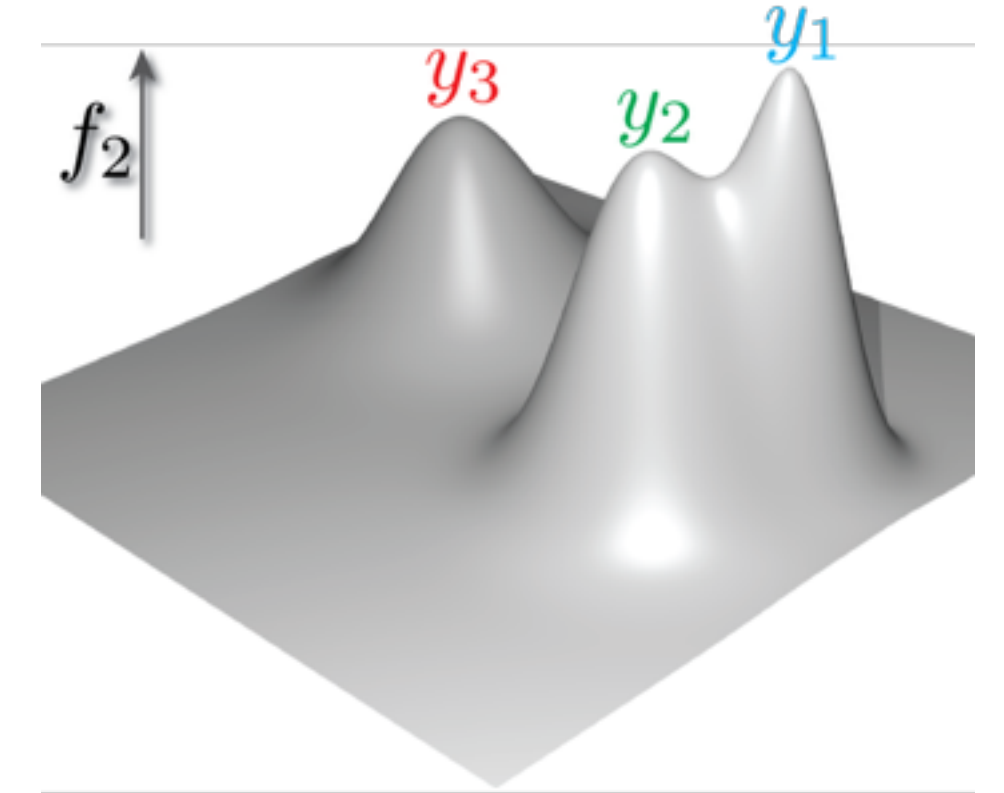
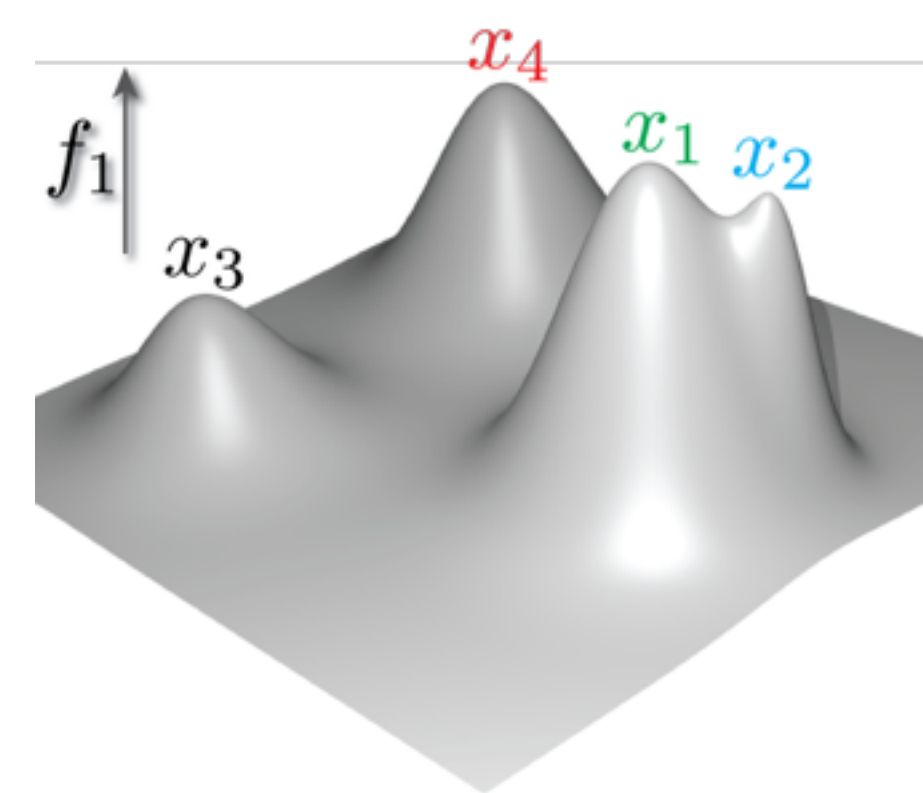
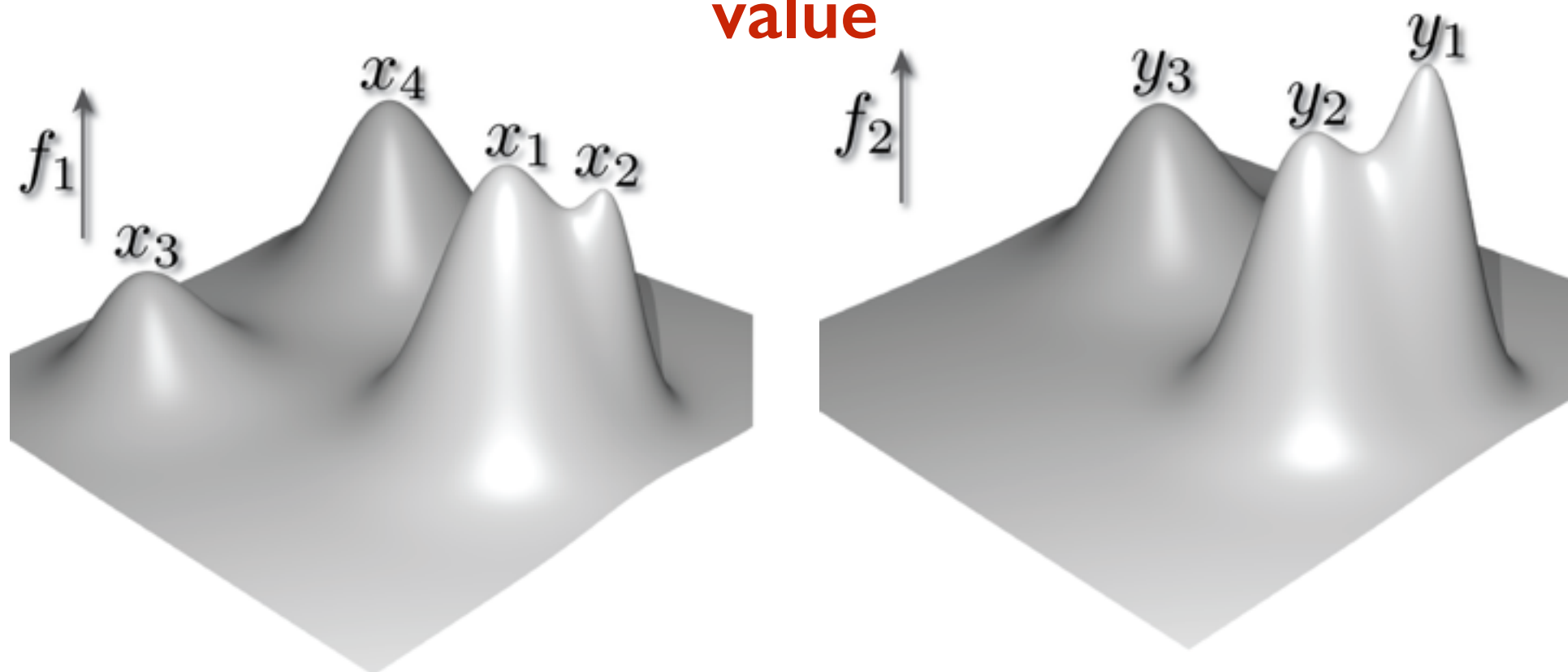
Is there a way to compute a **locality-aware** similarity measure?

# Maximum Topology Matching



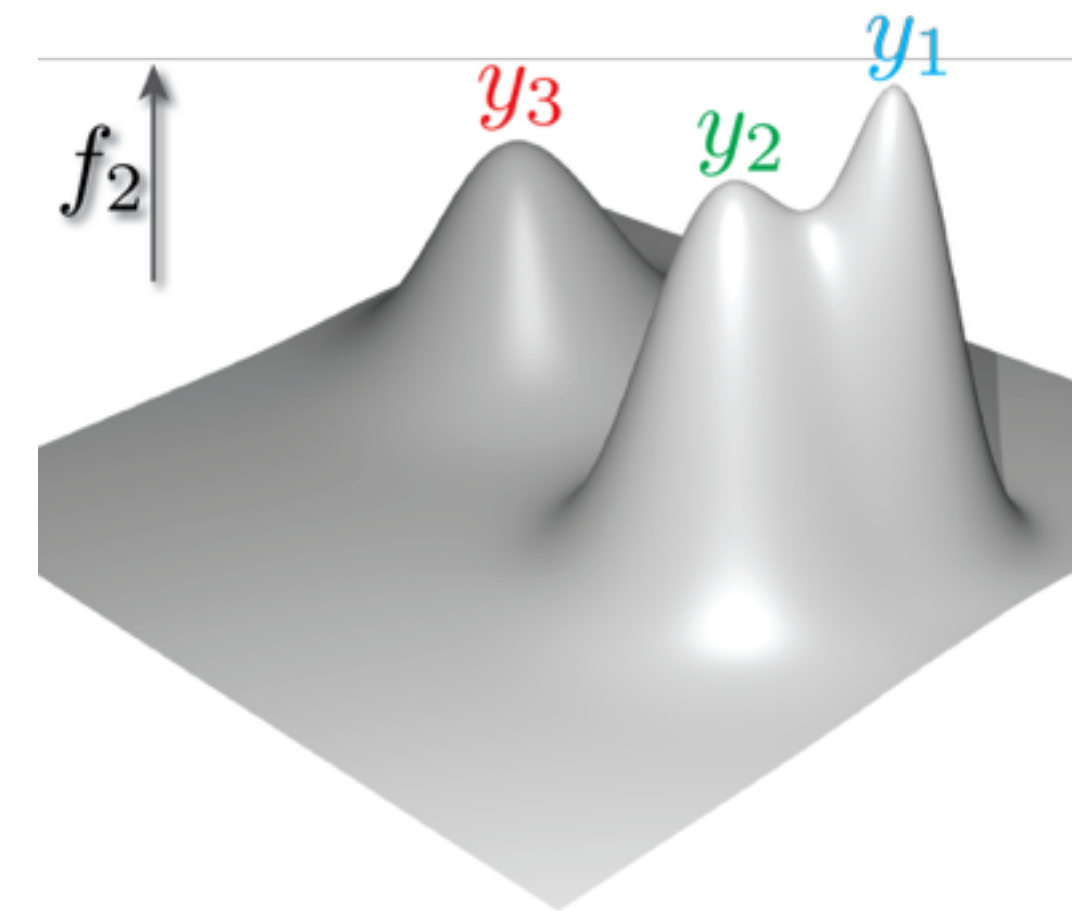
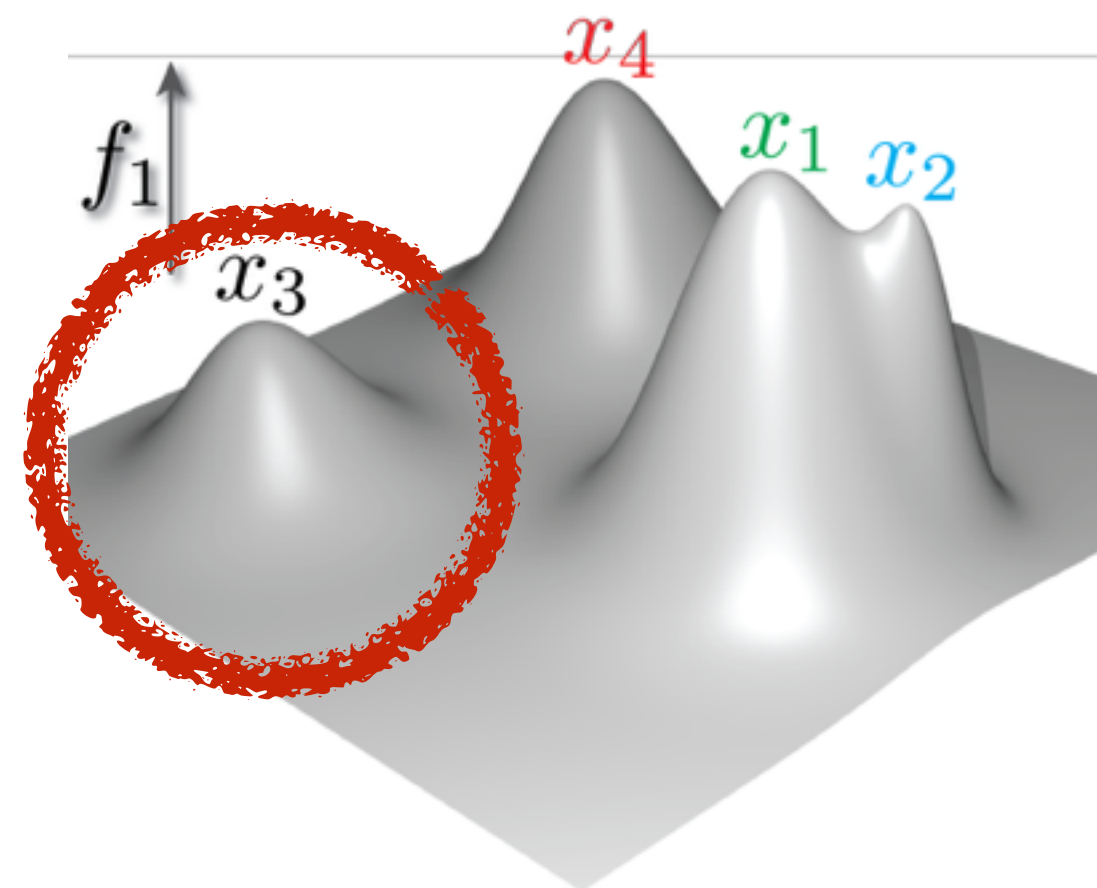
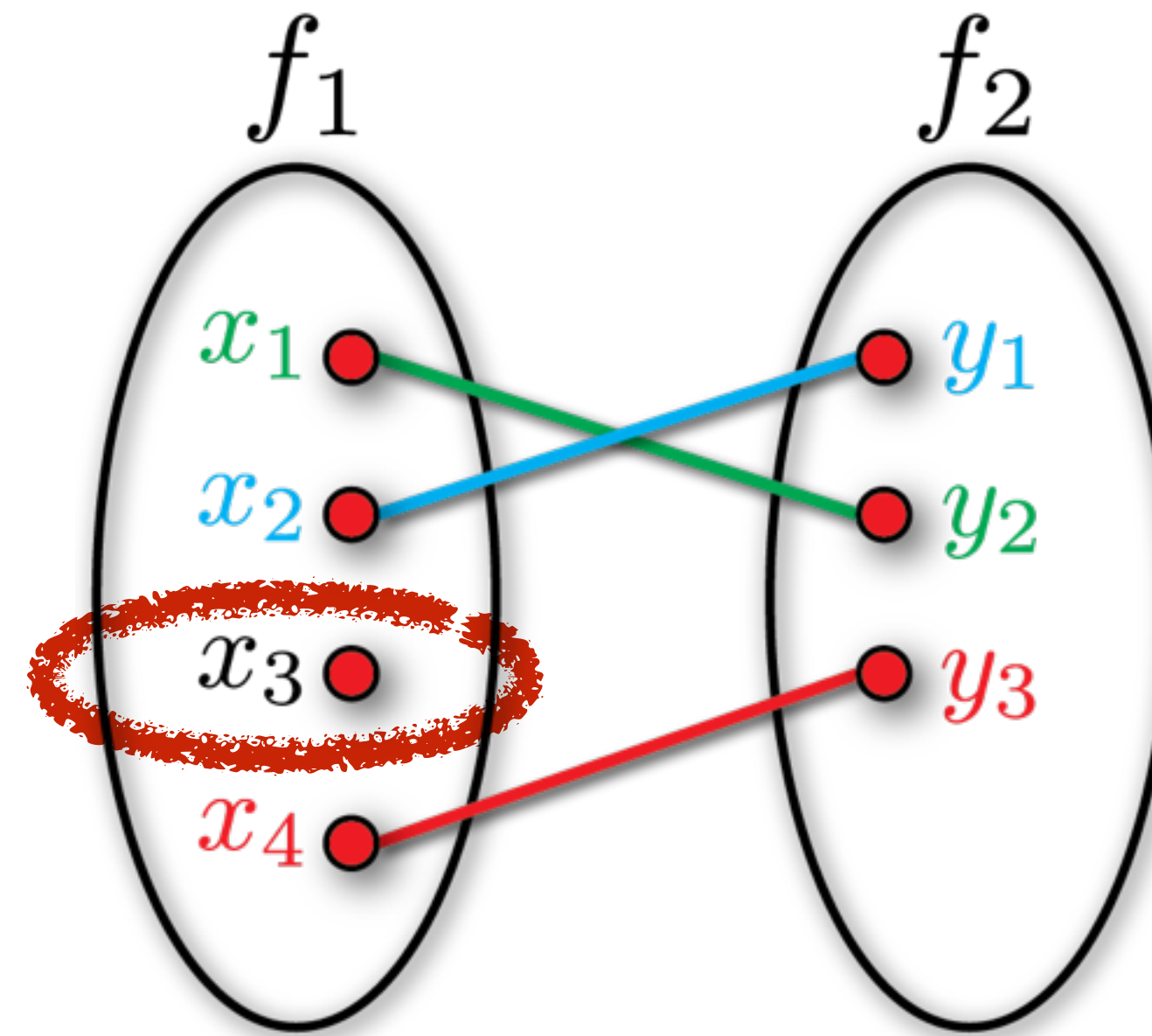
$$w_{a,b} = (1 - \delta_{a,b}) e^{\frac{d_g(a,b)}{r^2}}$$

**function value** **locality**



# Topological Similarity

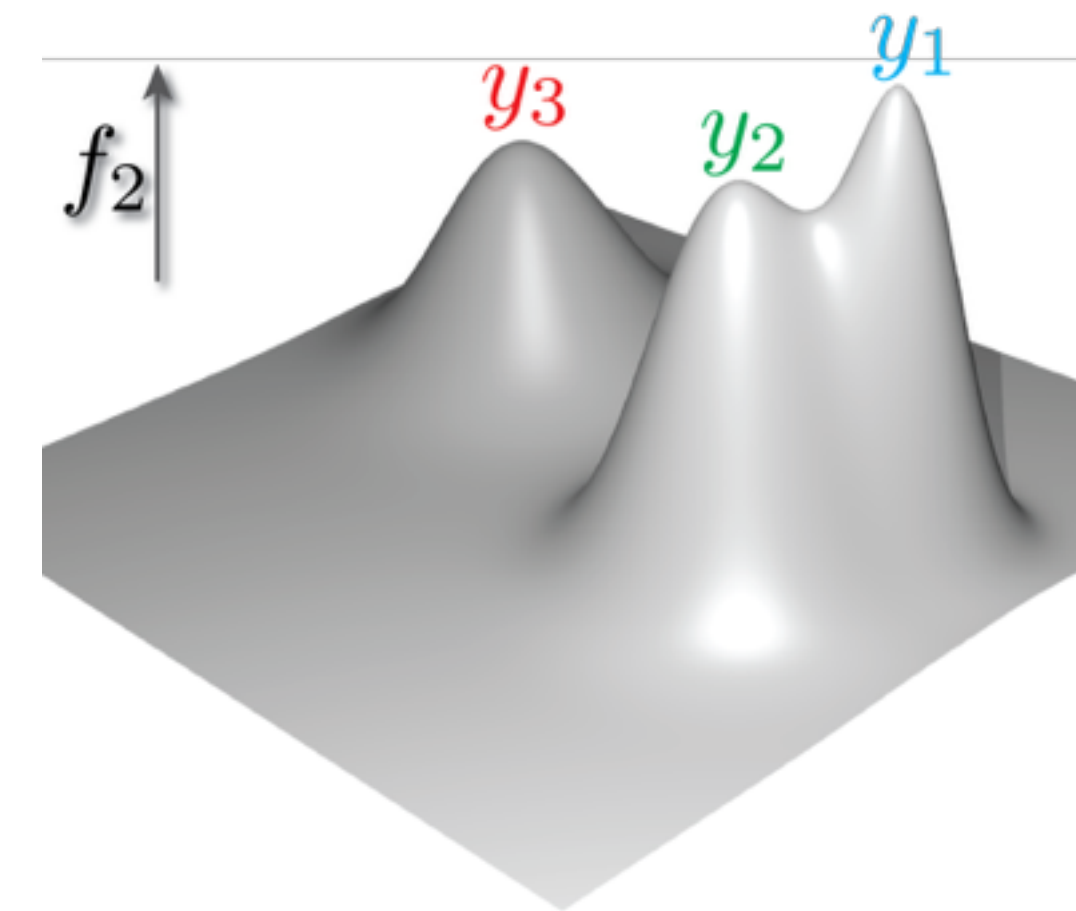
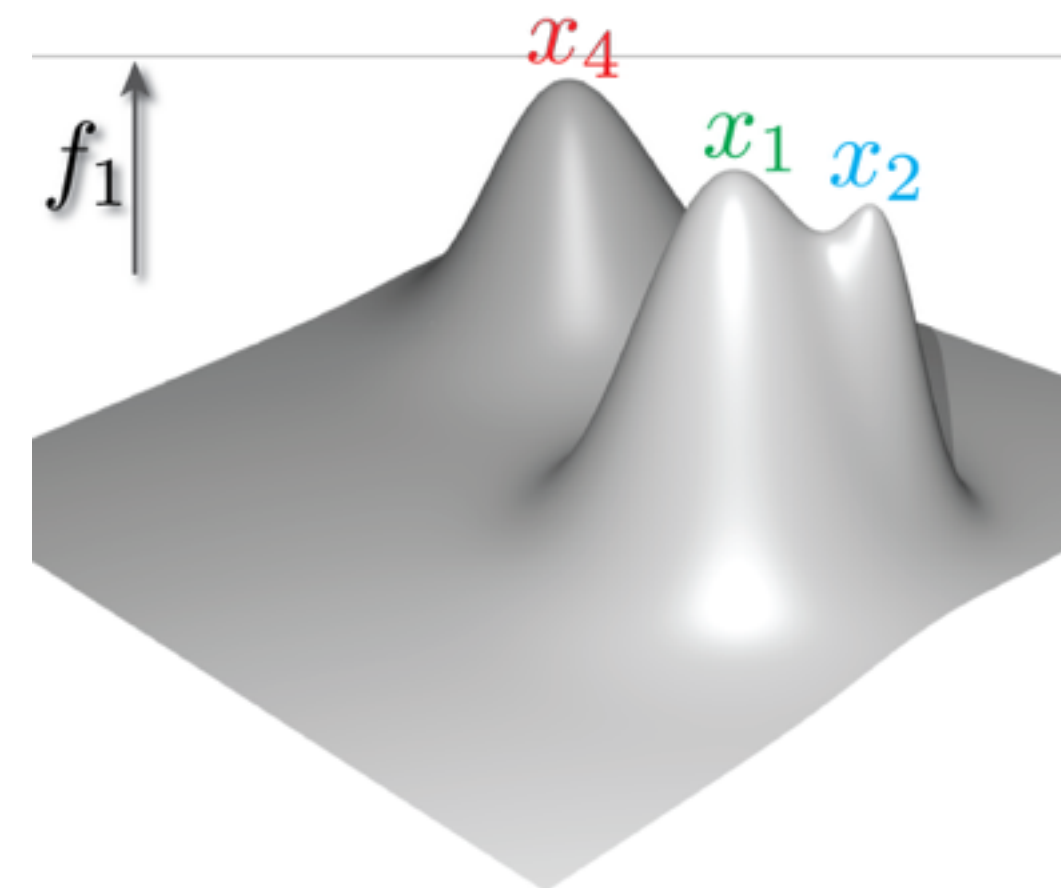
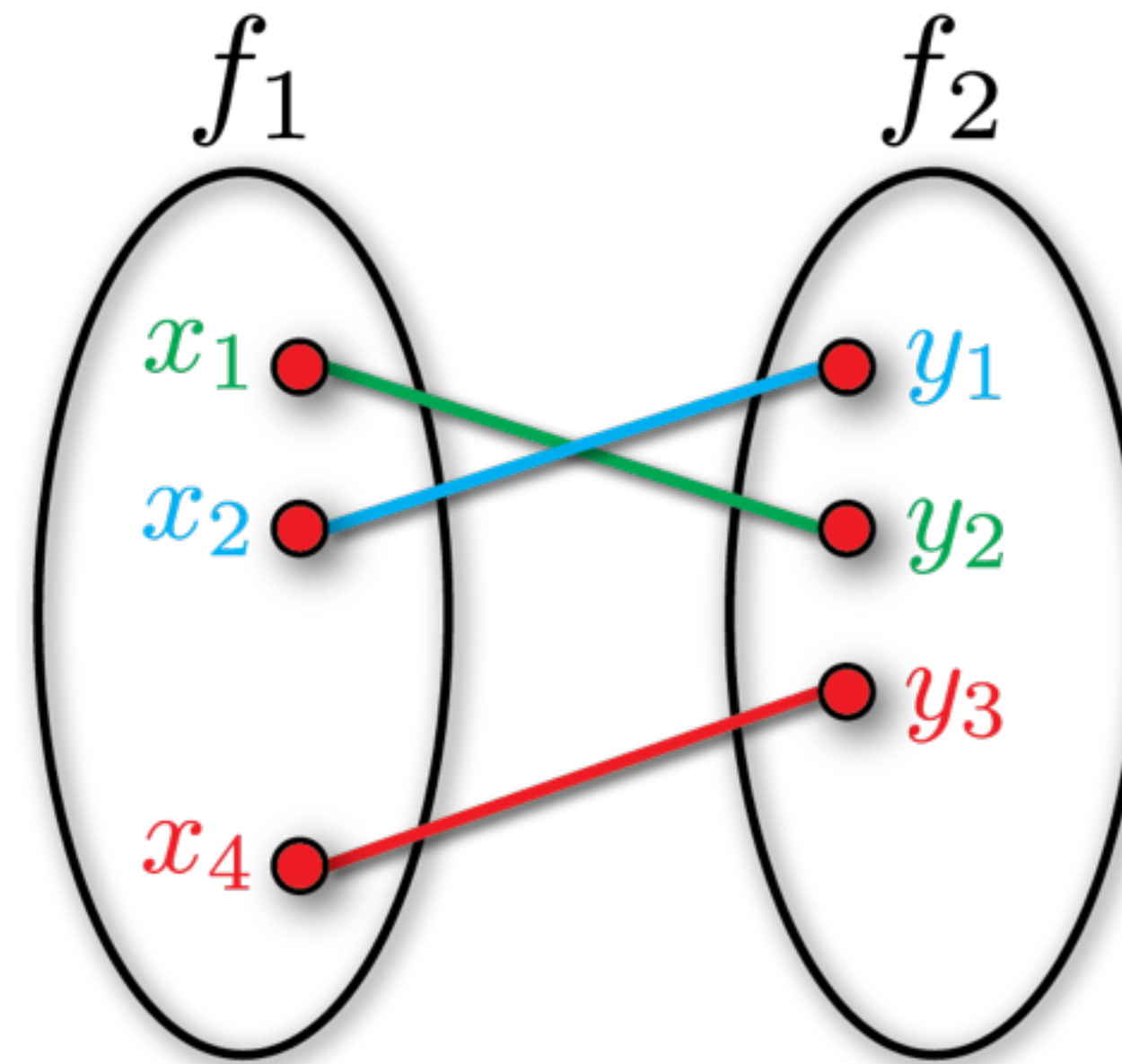
Intuition: It is the minimum simplification required to obtain a perfect matching between two functions





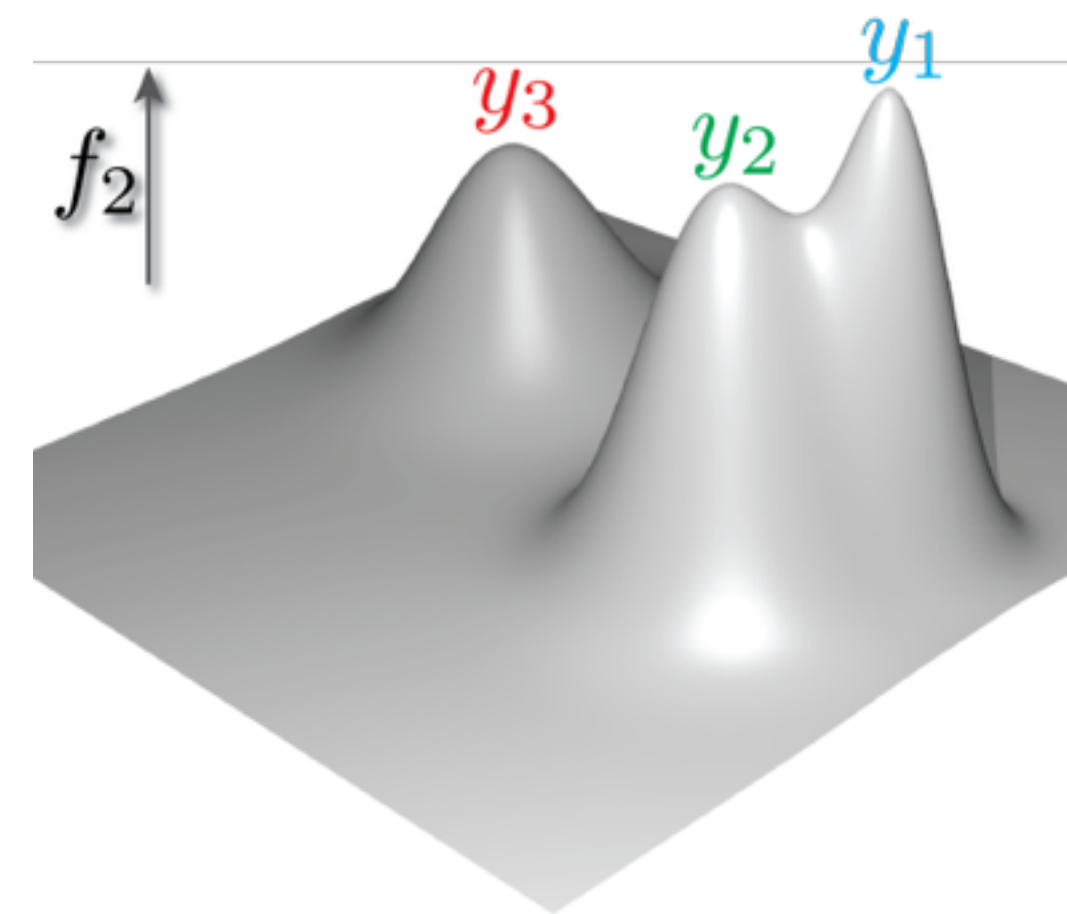
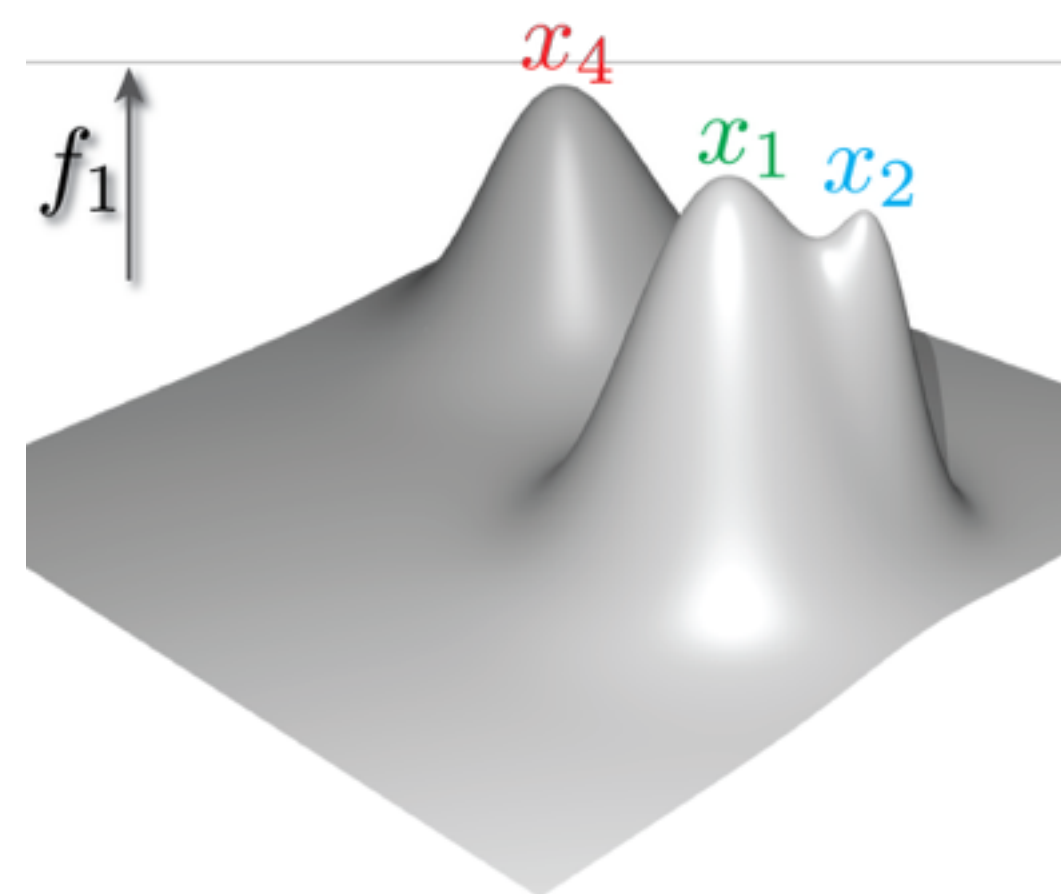
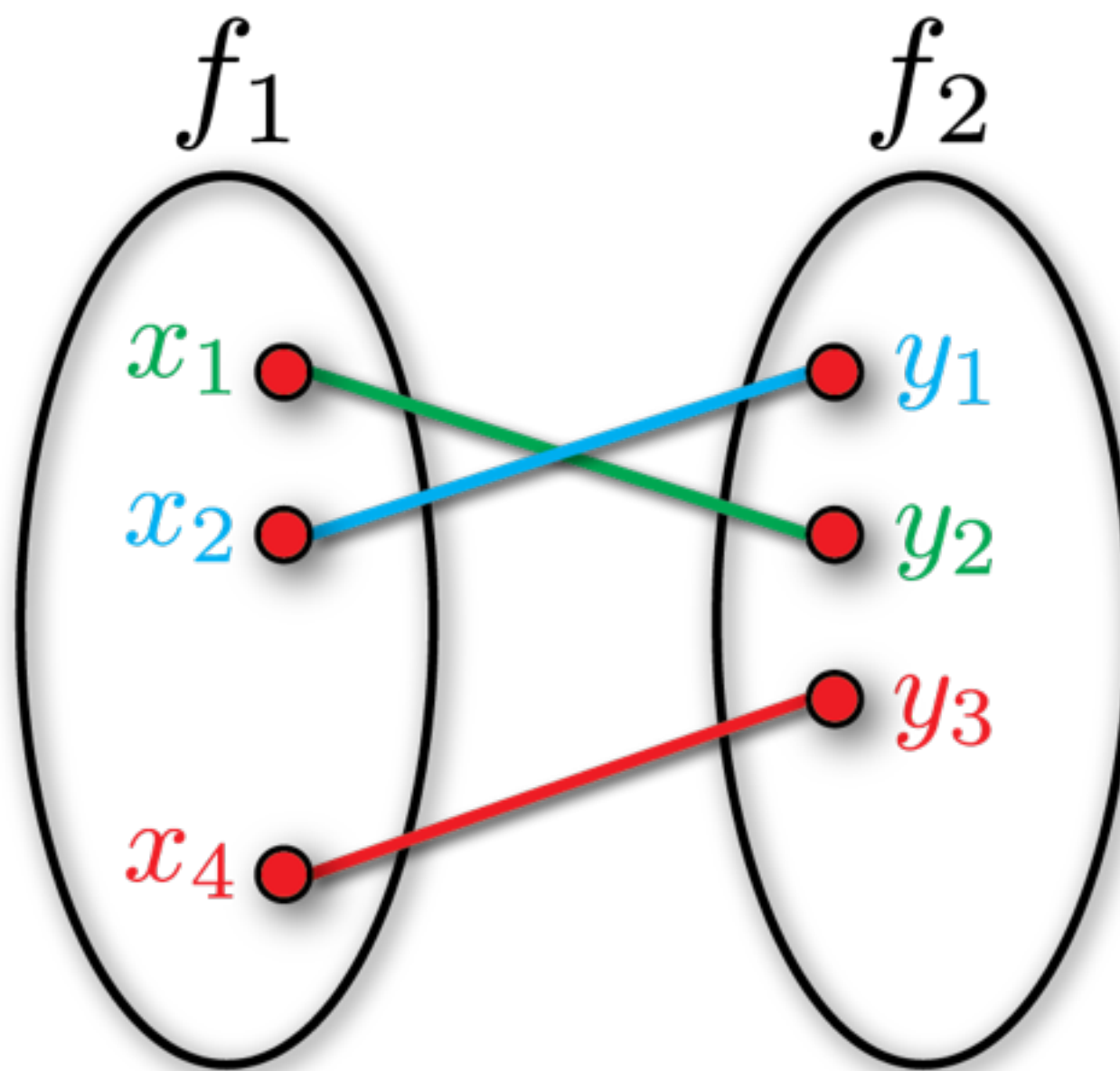
# Topological Similarity

Intuition: It is the minimum simplification required to obtain a perfect matching between two functions



# Functional Similarity

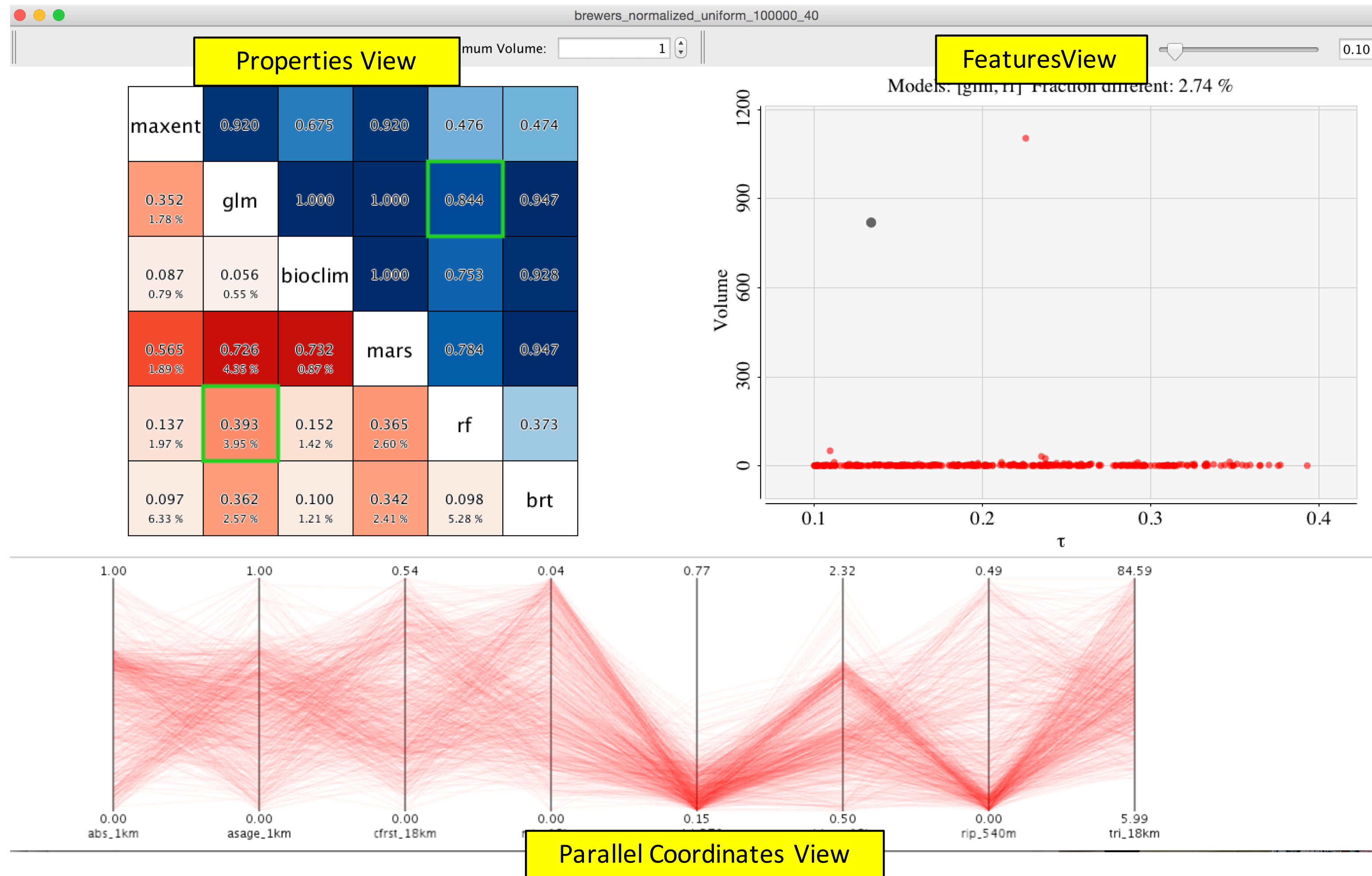
Intuition: Measures the amount of change required to get identical functions



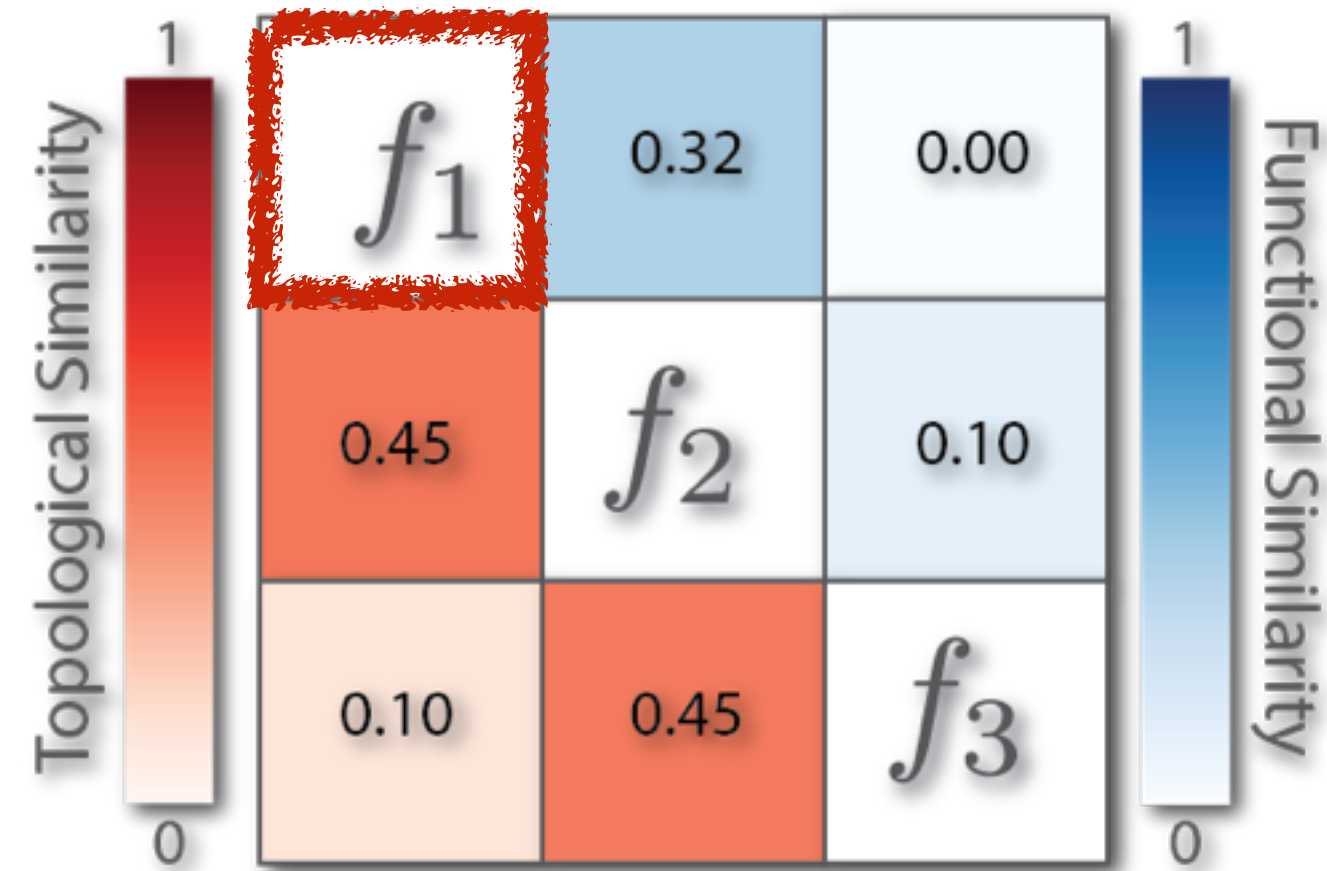
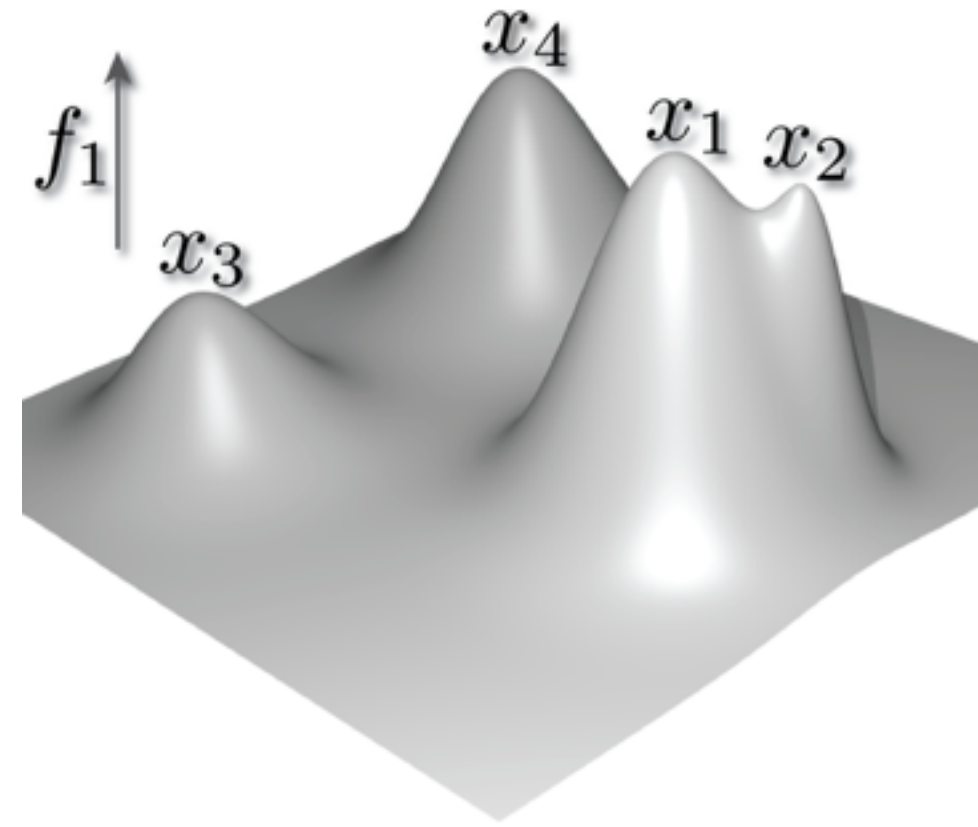
# Exploration Framework



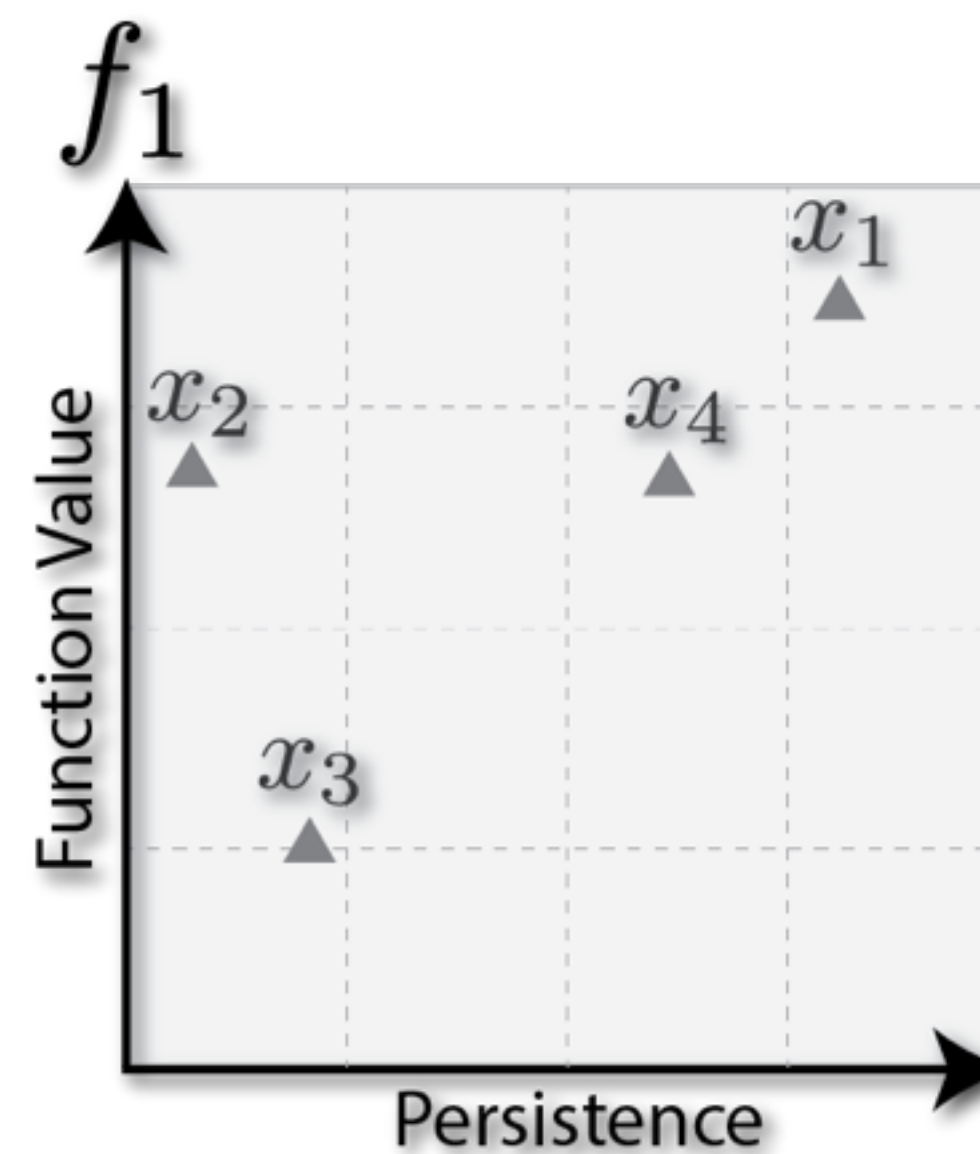
# Exploration Framework



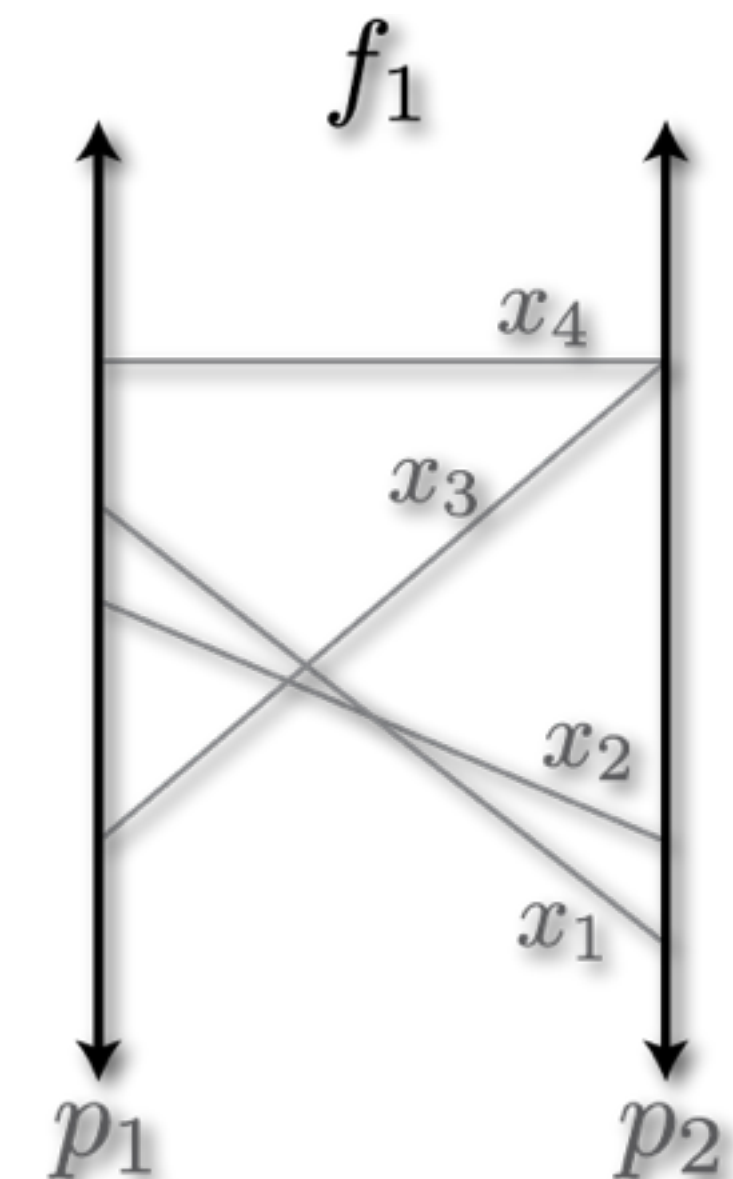
# Exploring a SDM



Properties View

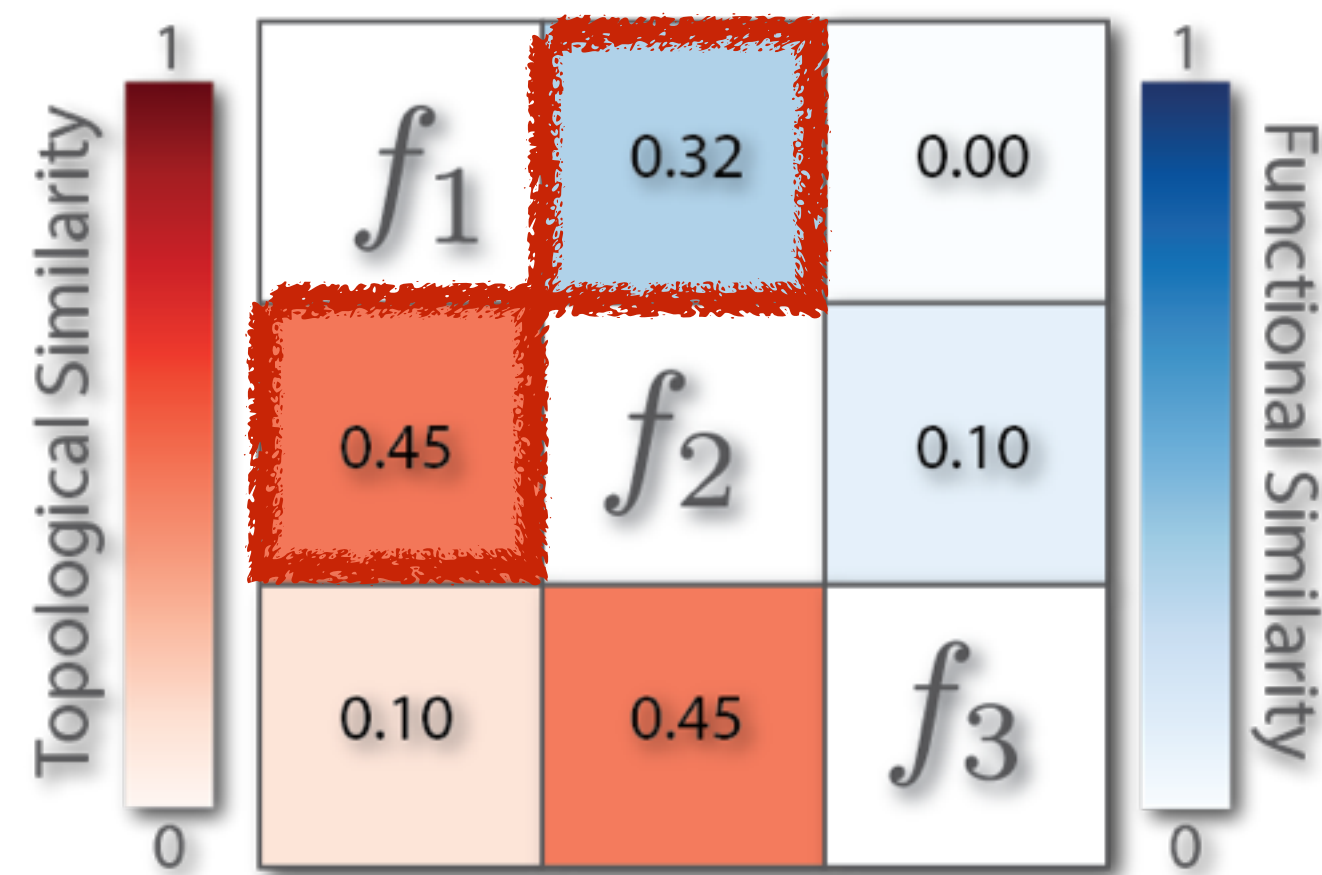
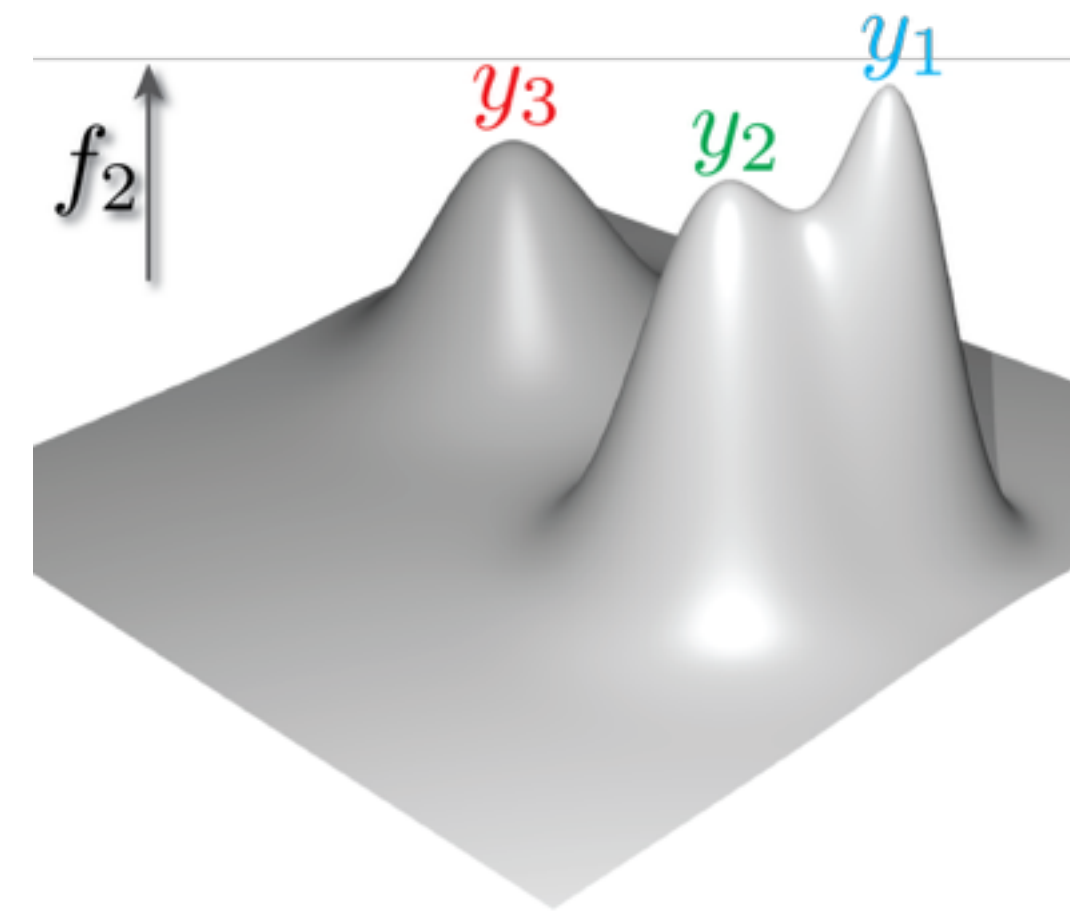
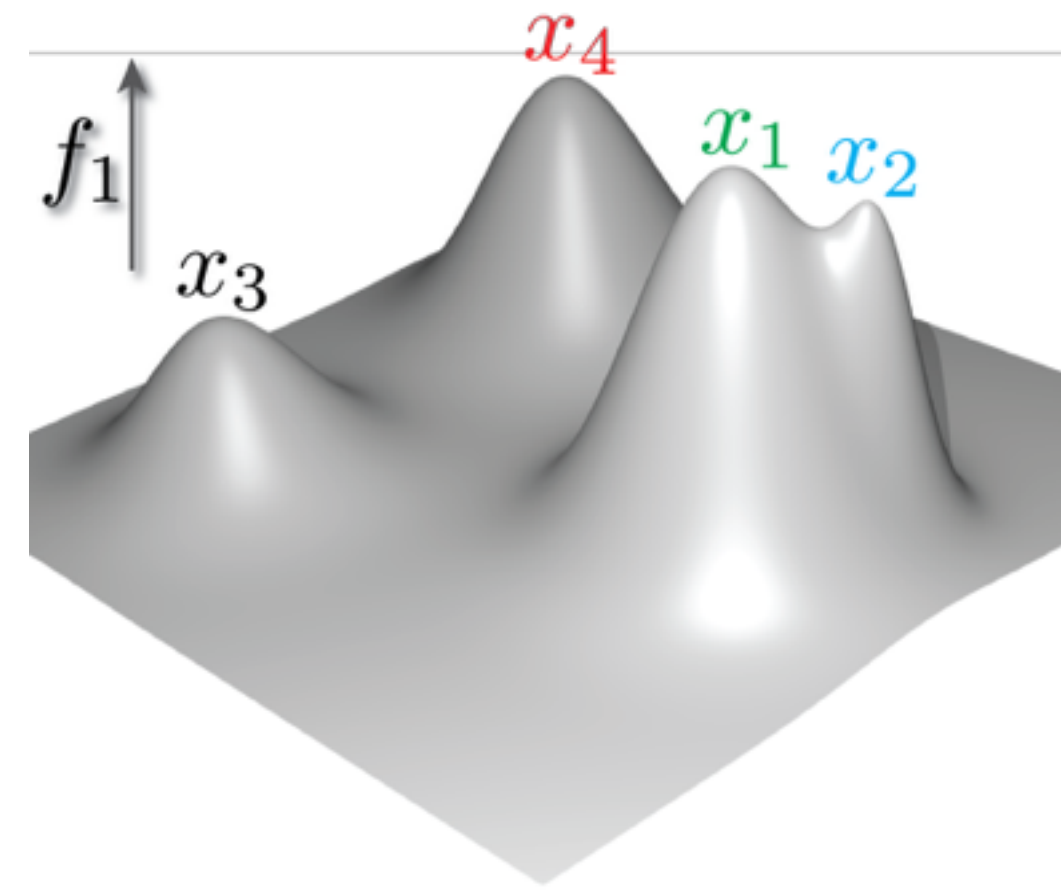


Features View

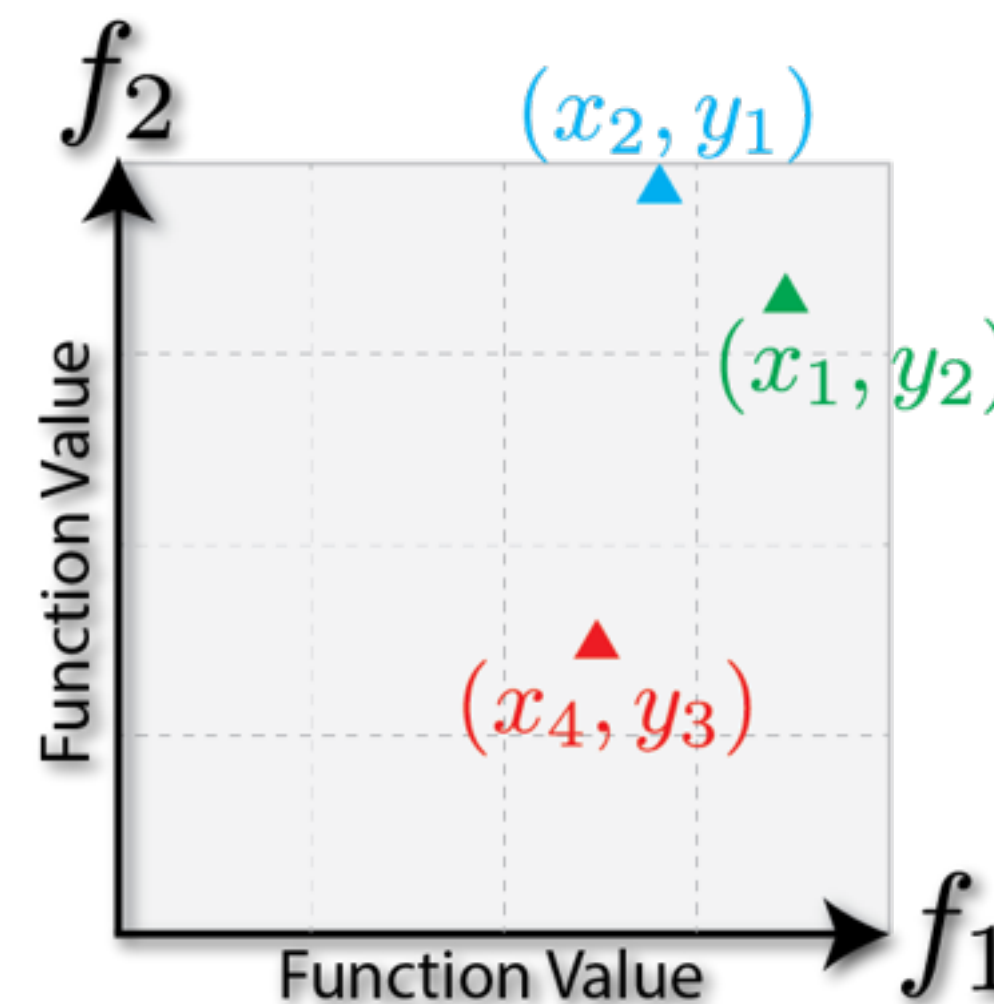


Parallel Coordinates View

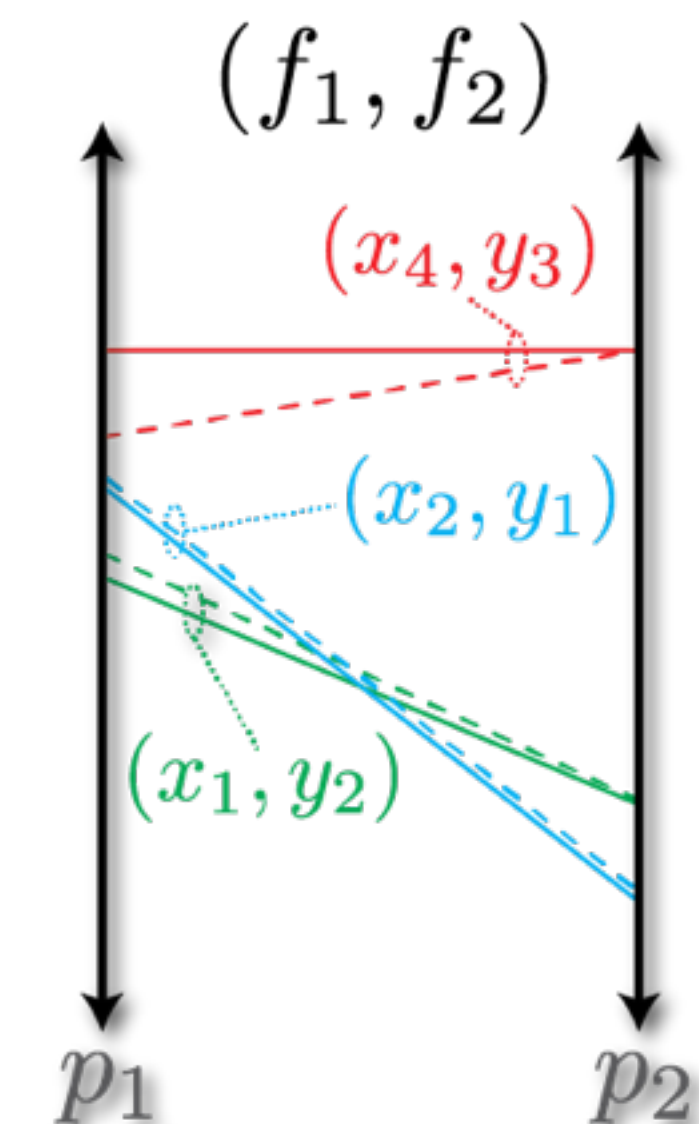
# Similarities Between Models



Properties View



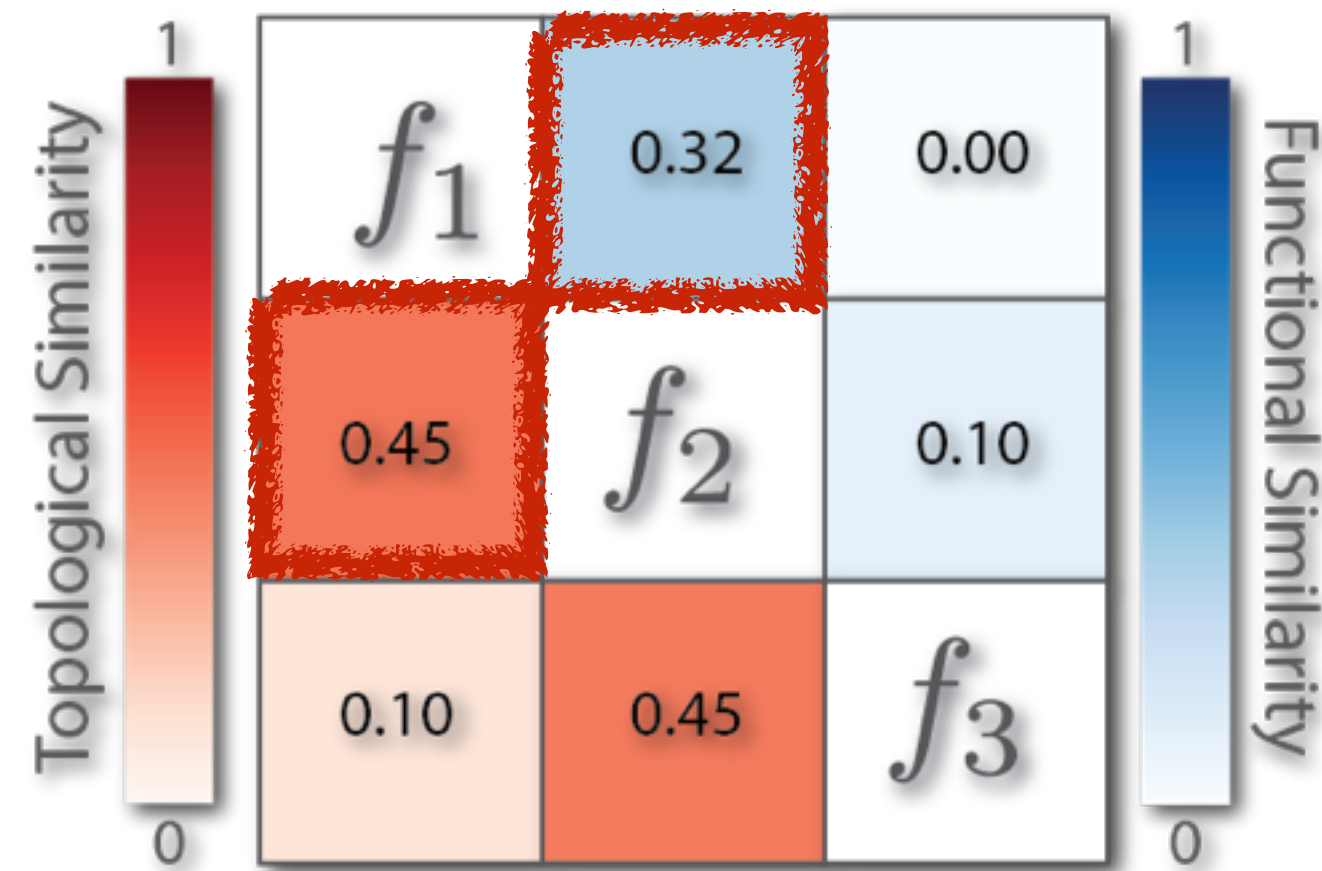
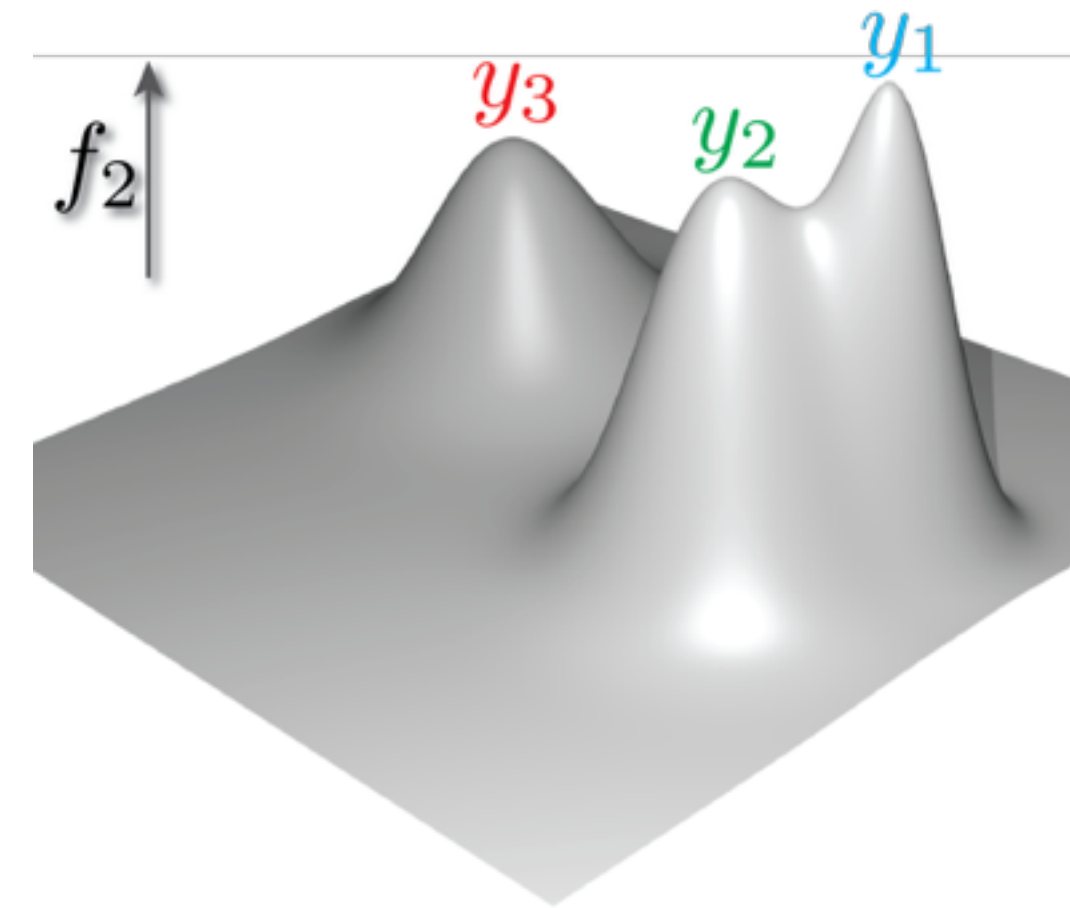
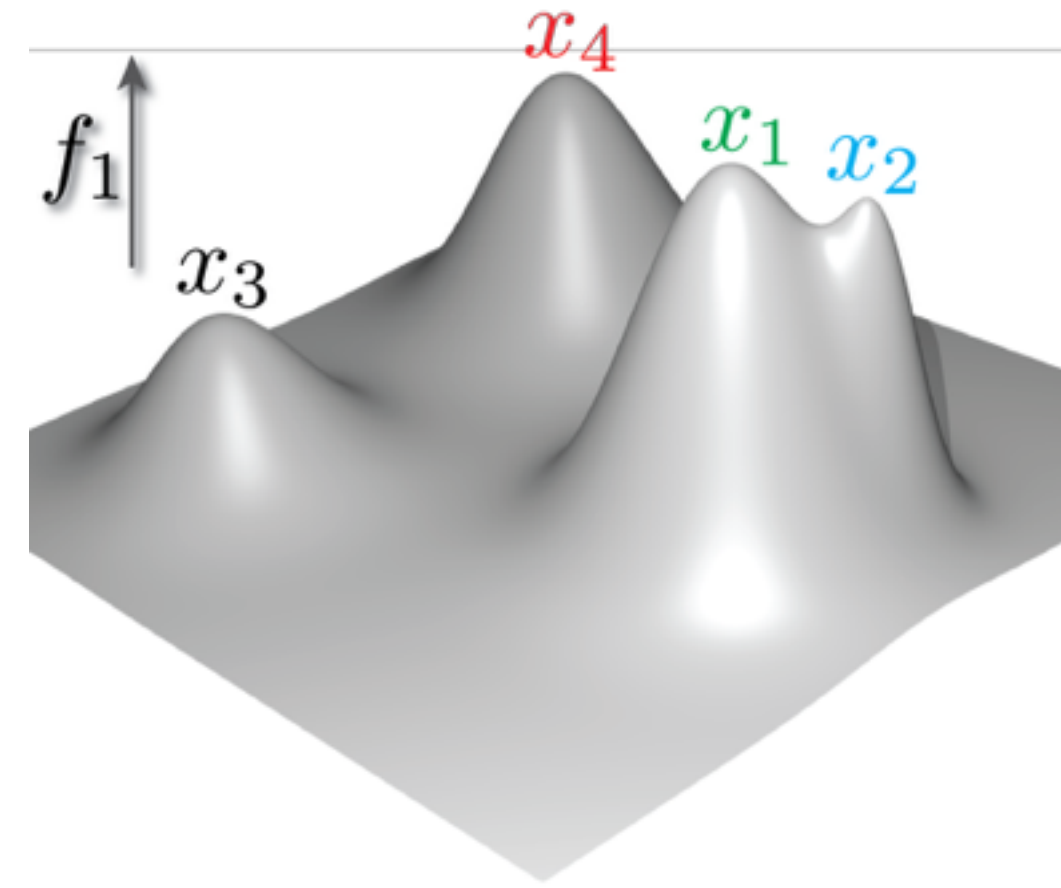
Features View



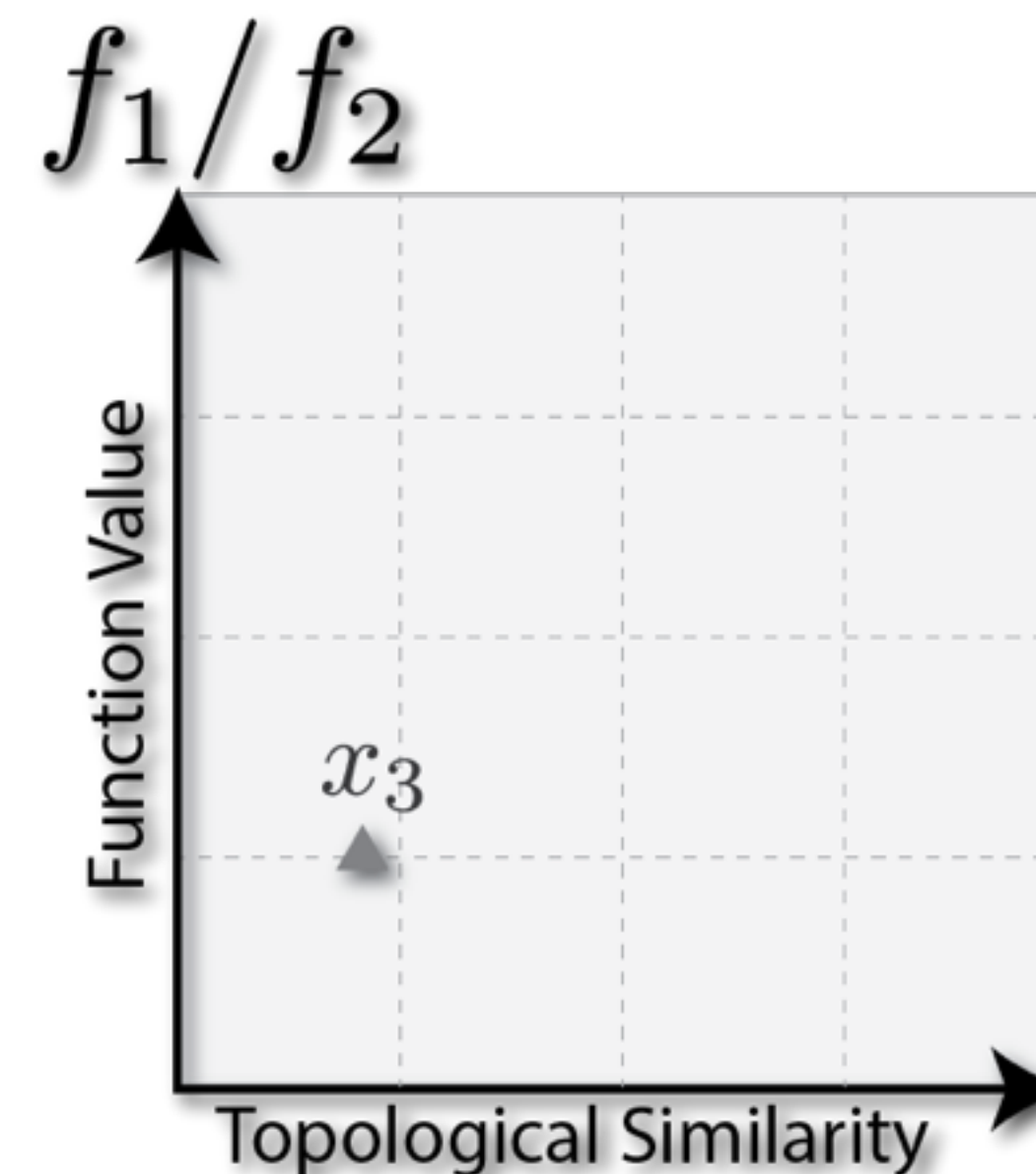
Parallel Coordinates View



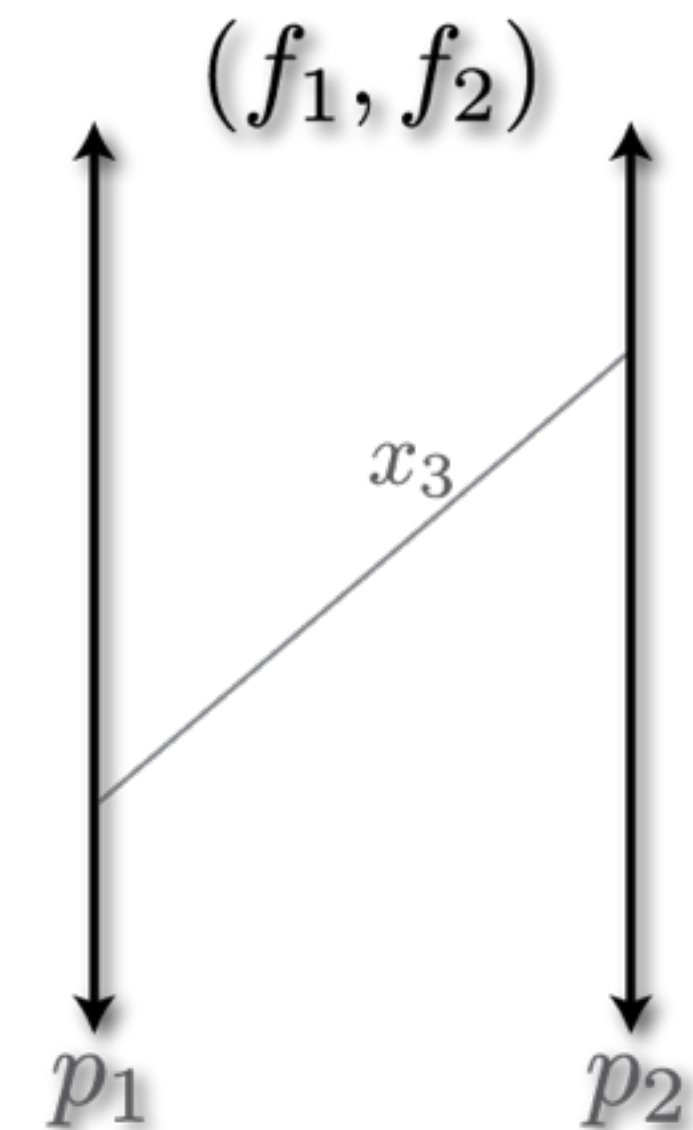
# Differences Between Models



Properties View



Features View



Parallel Coordinates View

# Case Studies

# Data Sets

- Brewer's Sparrow
  - 8 predictors
  - 5 models
- Sagebrush
  - 8 predictors
  - 5 models
- Spruce Fir
  - 9 predictors
  - 5 models

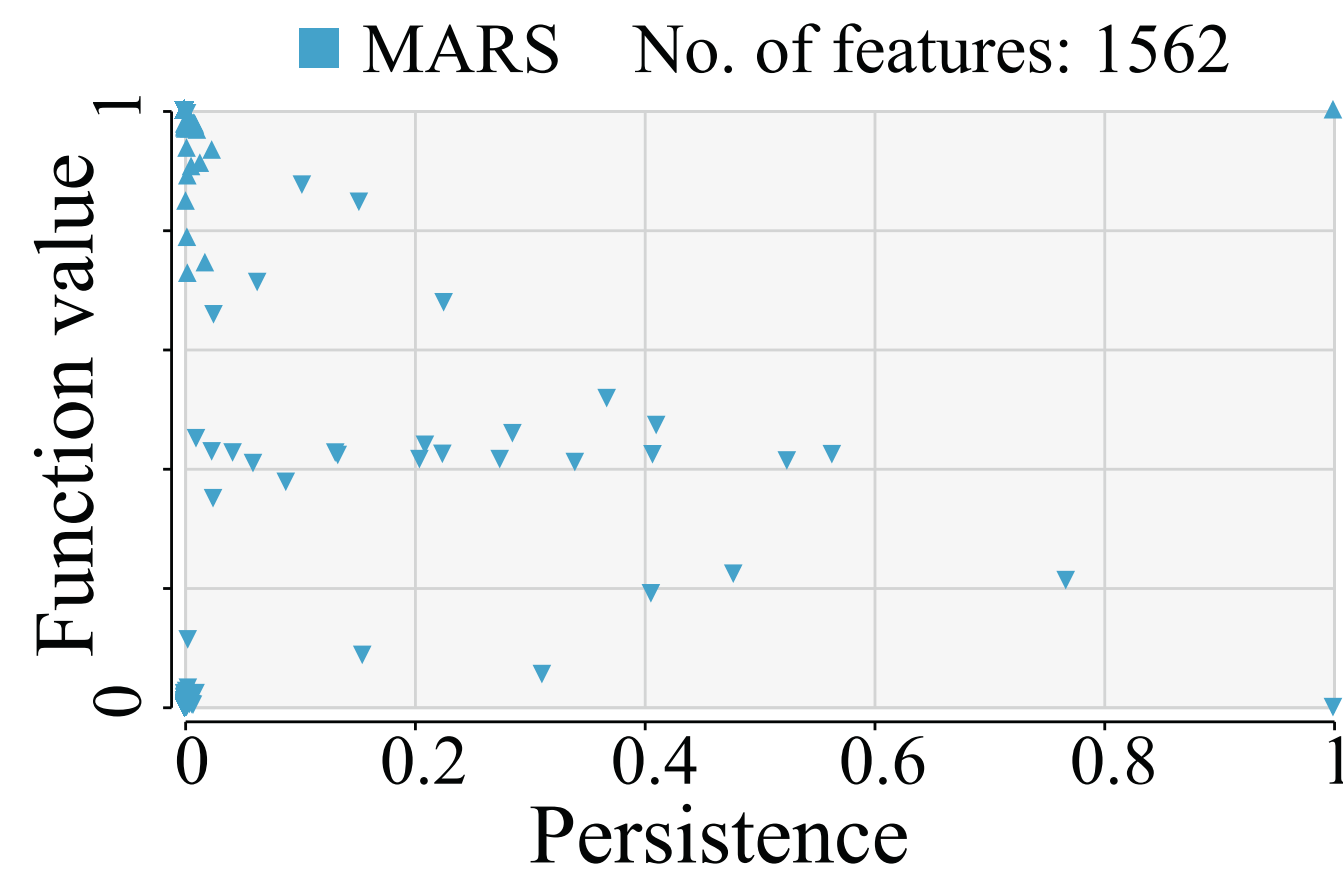


# Implementation

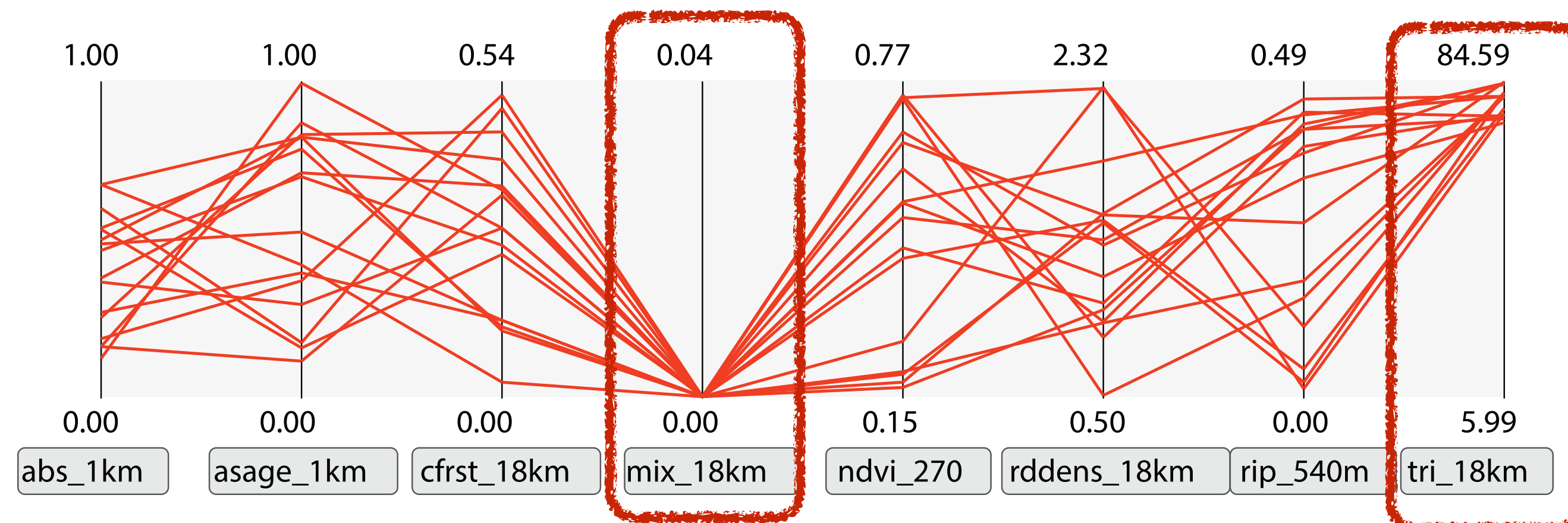
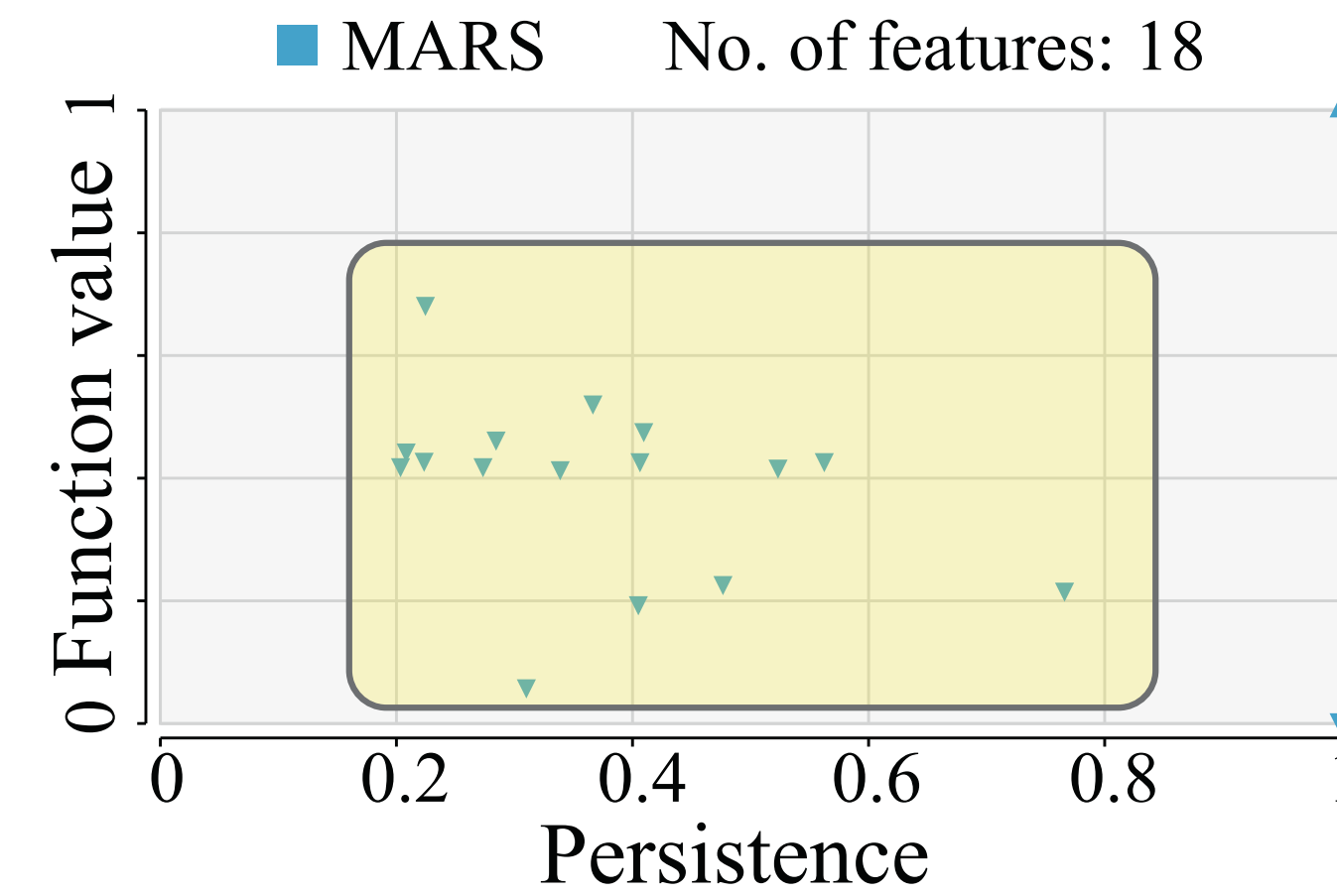
- Domain of SDM
  - Discretize as a graph
  - Sample  $n=10^5$  points in the high-dimensional space
    - Latin Hypercube Sampling
  - Compute the k-nearest-neighbor graph ( $k=40$ ) for each SDM
- Scalar function
  - Function values are computed using the SDM on the vertices of the Graph
  - Linearly interpolated within each edge
- Cut-off radius for computing edge weights  $r = 0.1$



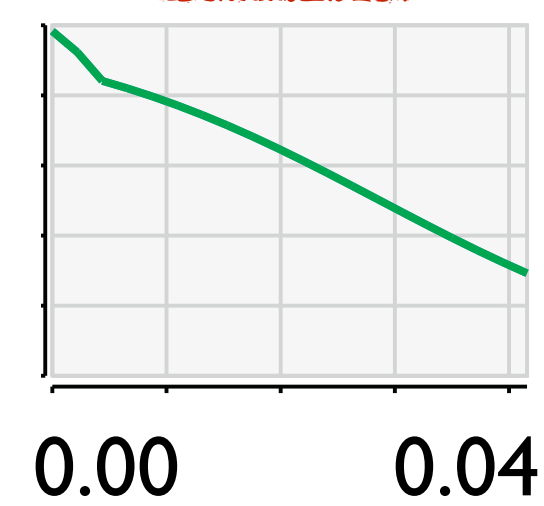
# Exploring a SDM



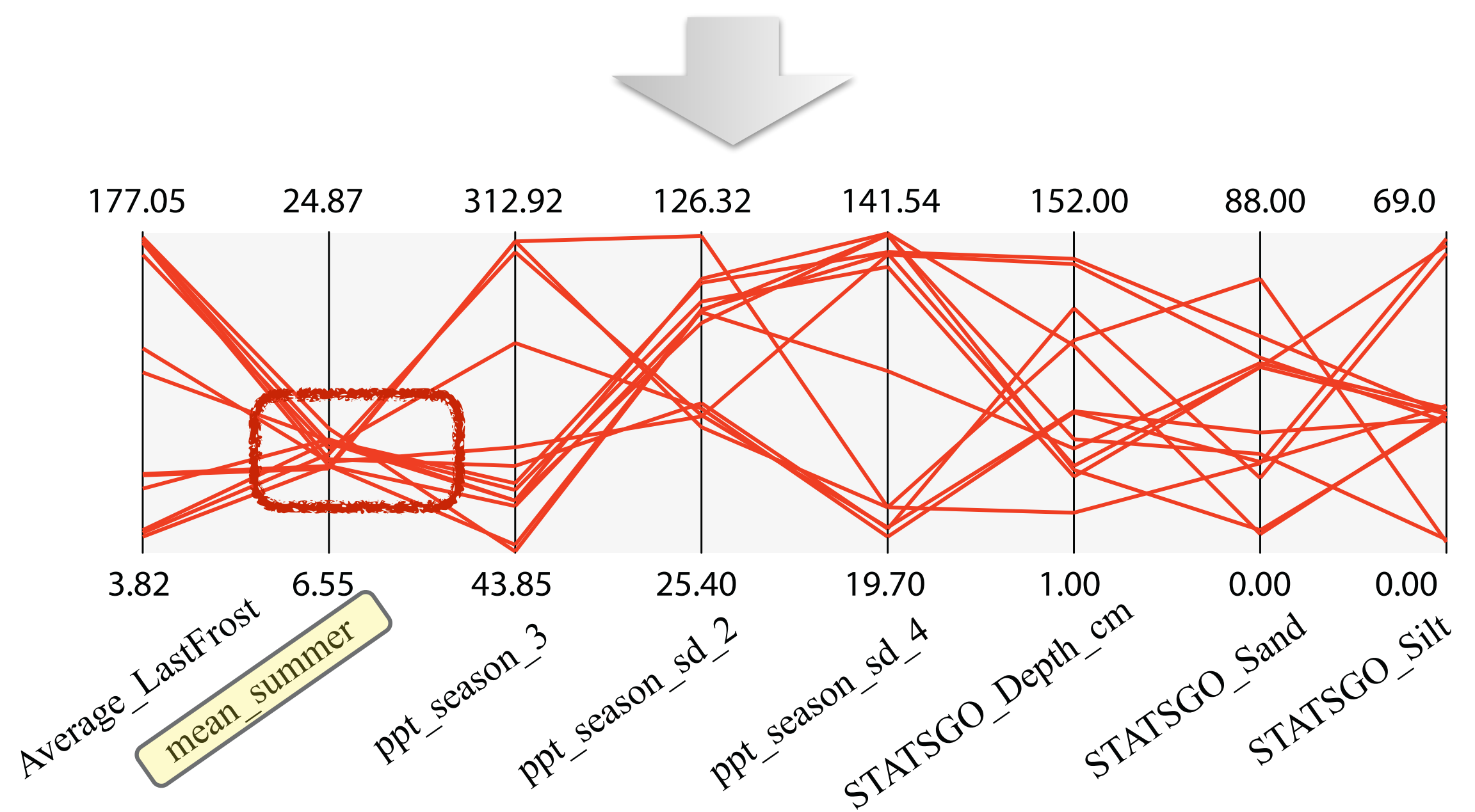
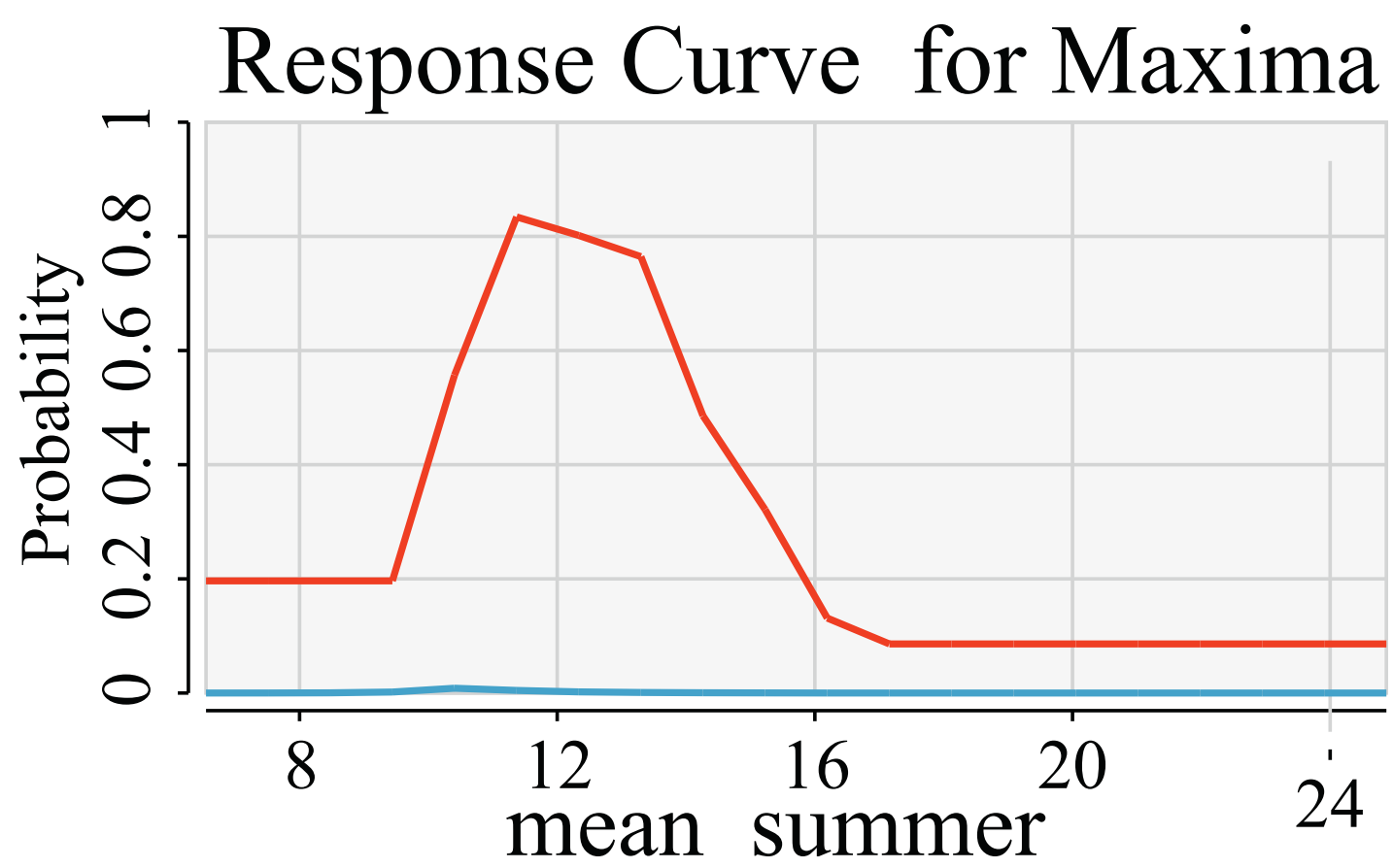
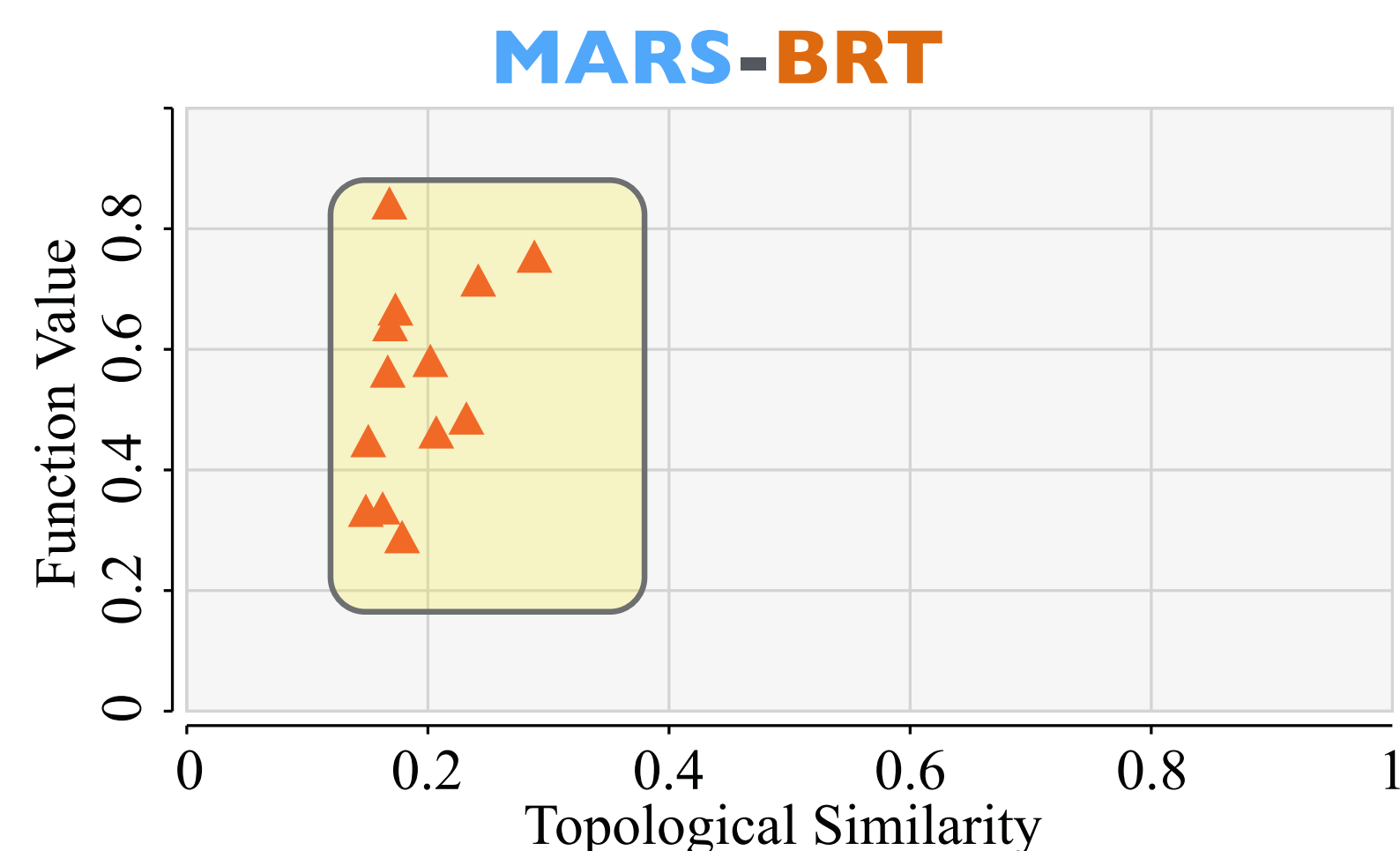
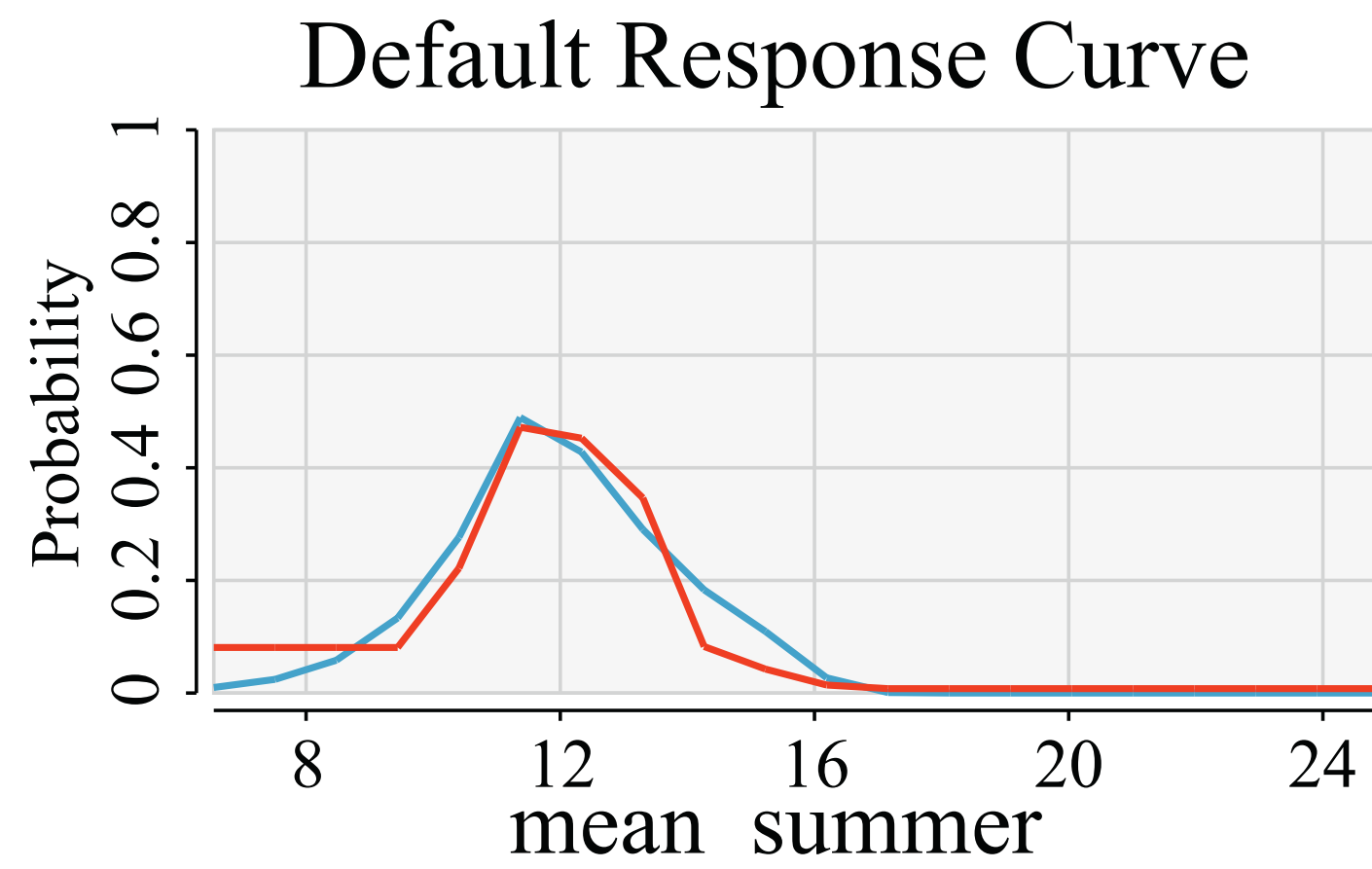
➔  
Simplification



default response curve ➔



# Exploring Differences: MARS vs BRT



# Conclusions

- Positive feedback from Ecologists
  - Surprised by results
  - Integrating into SAHM package for VisTrails
- Other Contributions
  - Robustness to noise
    - perturb function values
    - perturb extrema locations
  - Experimentally evaluate effect of parameters to the similarity measures
    - sample size
    - # neighbors
    - neighborhood radius  $r$

# Future Work

- Each dimension normalized between 0 and 1
  - Can they be standardized instead?
- Neighborhood radius fixed to 0.1 based on discussions with our collaborators
  - Can a different weighting scheme be used irrespective of the domain?
- Use other metaphors to visualize SDMs
  - Eg. topological spines



# Acknowledgements

This work was supported in part by

- Google Faculty Award
- IBM Faculty Award
- Moore-Sloan Data Science Environment at NYU
- NYU School of Engineering
- NYU's Center for Urban Science and Progress (CUSP)
- AT&T
- NSF award CNS-1229185
- DOE
- NASA Biodiversity Program award NNH11AS091

# Using Maximum Topology to Explore Differences in Species Distribution Models

More Information:

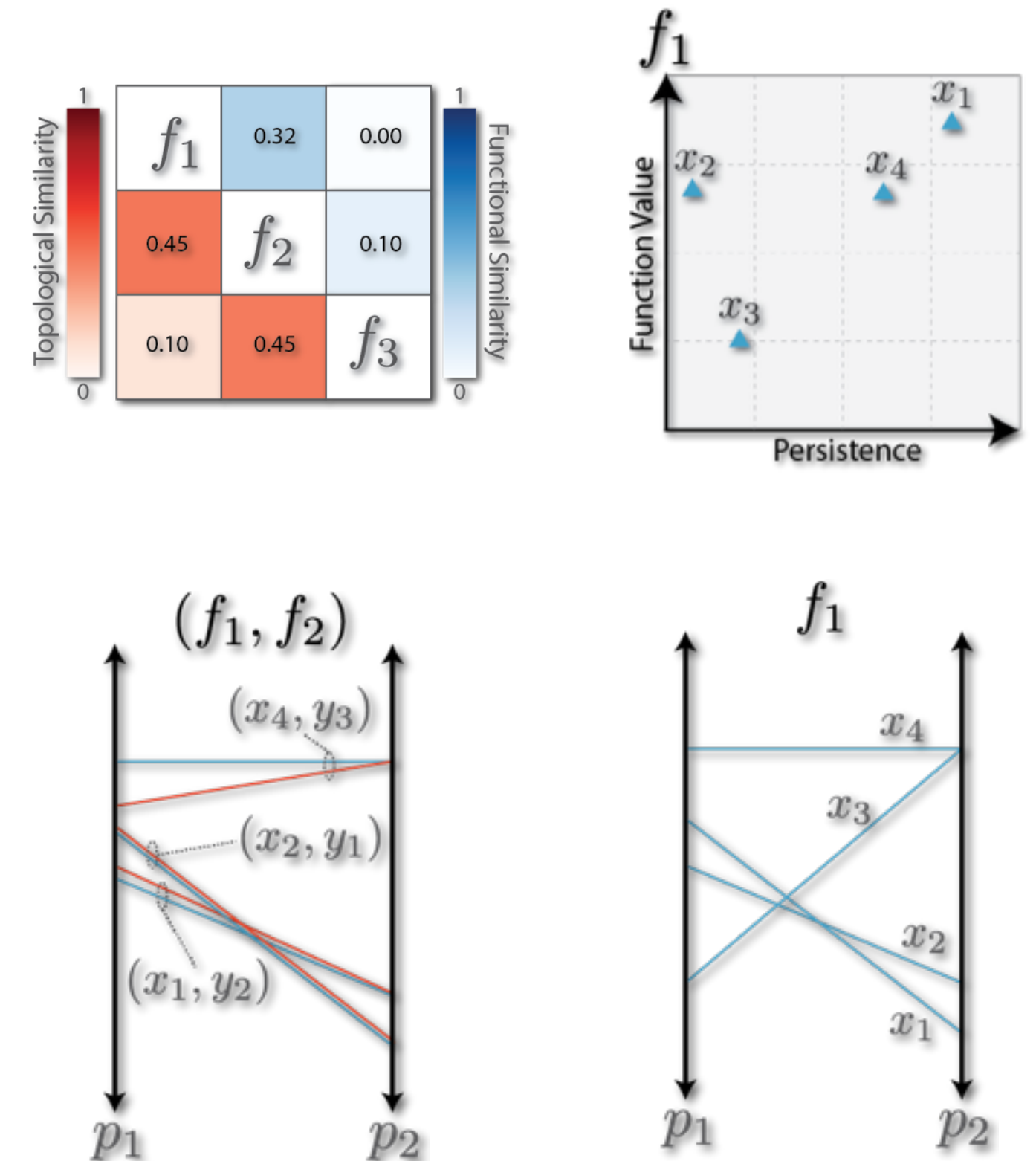
**Jorge Poco**

[jpocom@uw.edu](mailto:jpocom@uw.edu)

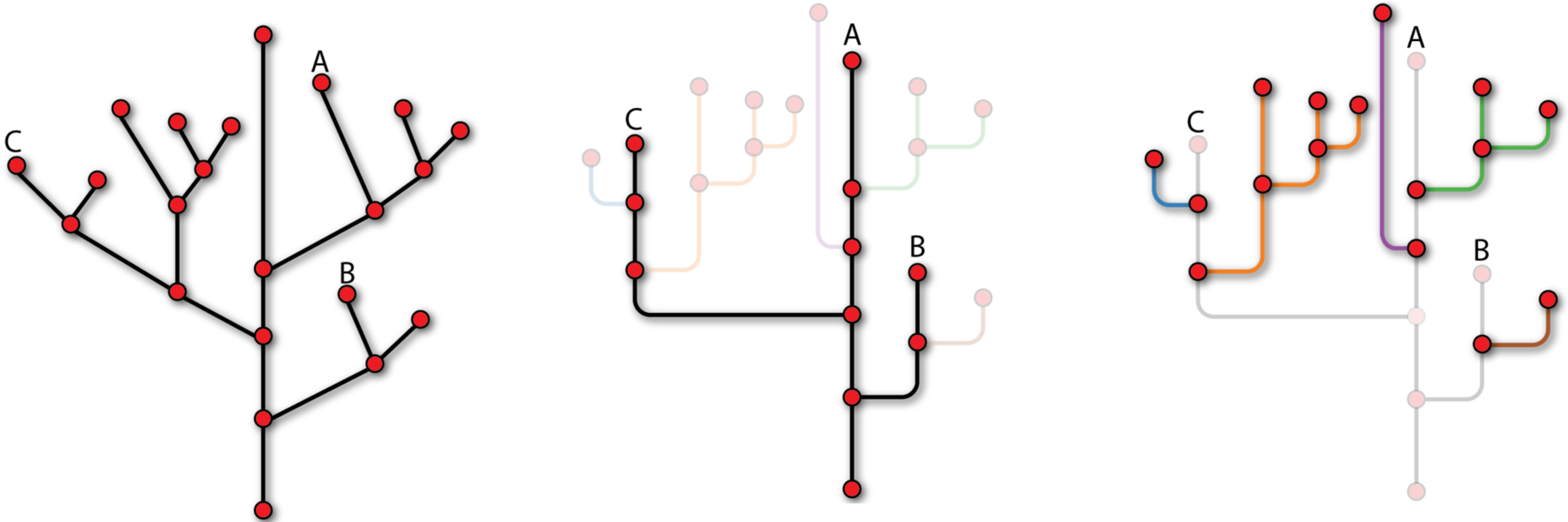
**Harish Doraiswamy**

[harishd@nyu.edu](mailto:harishd@nyu.edu)

## Thank You!



# Similarity Measures: Topological Similarity



$$\tau = \max(\tau_1, \tau_2)$$

- Intuition: It is the minimum simplification required to obtain a perfect matching between two functions



# Effect to Noise