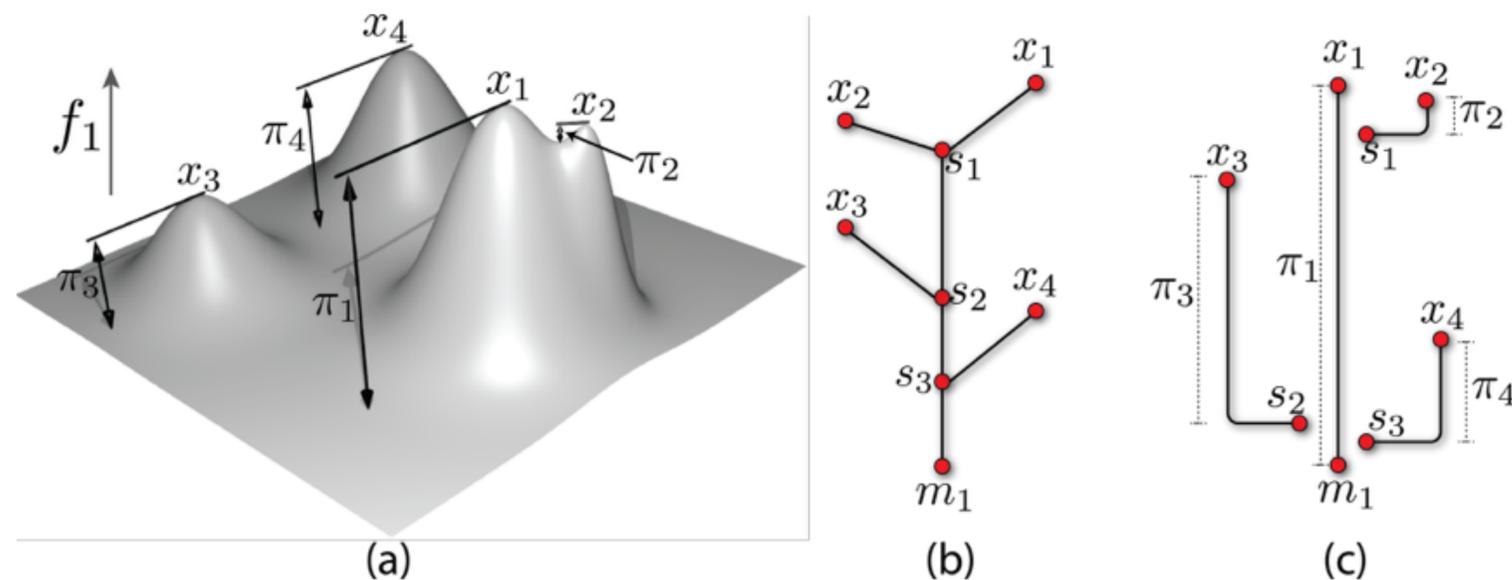


Using Maximum Topology Matching to Explore Differences in Species Distribution Models



Jorge Poco
Harish Doraiswamy
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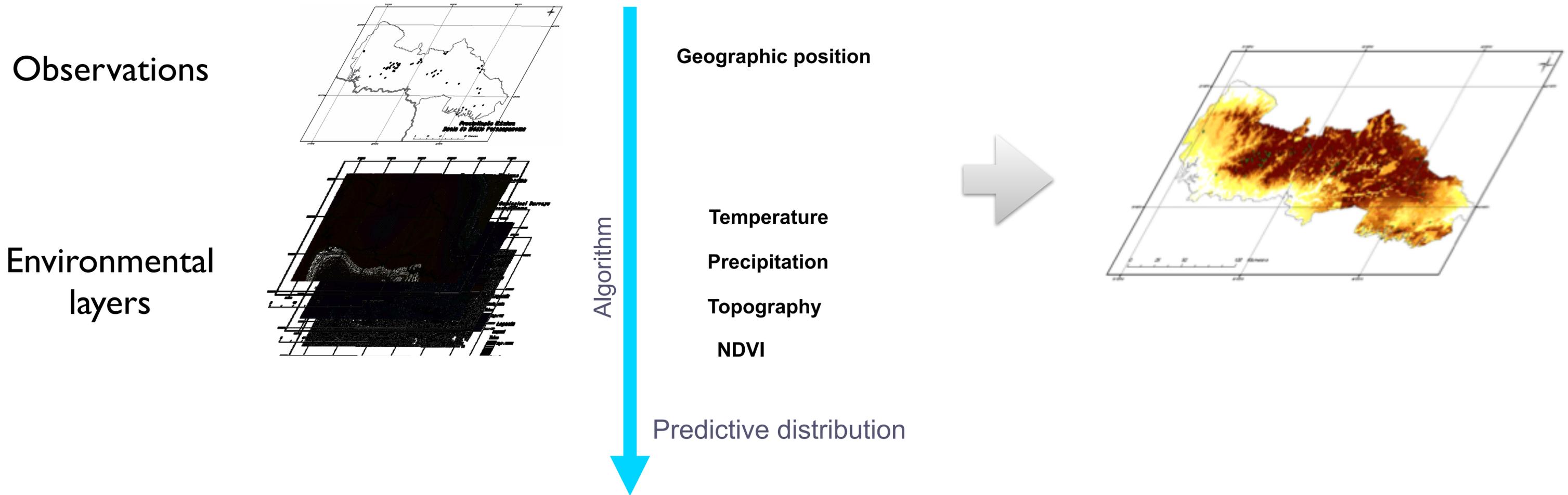


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Species Distribution Models (SDM)

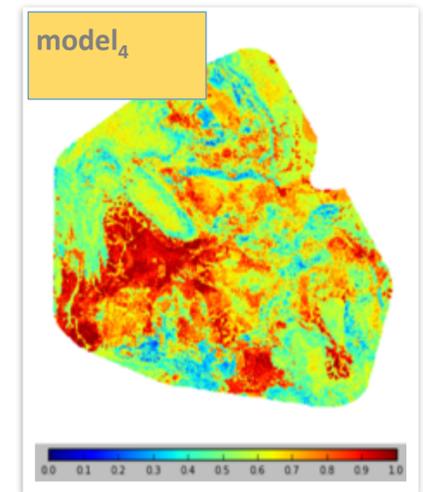
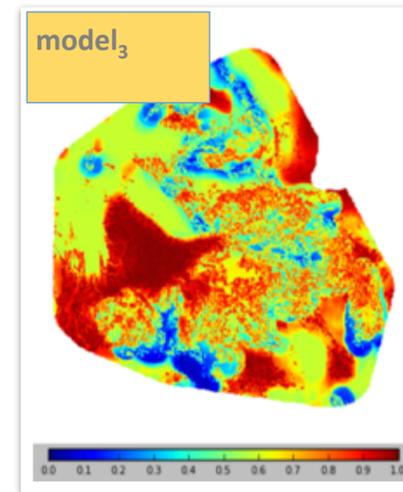
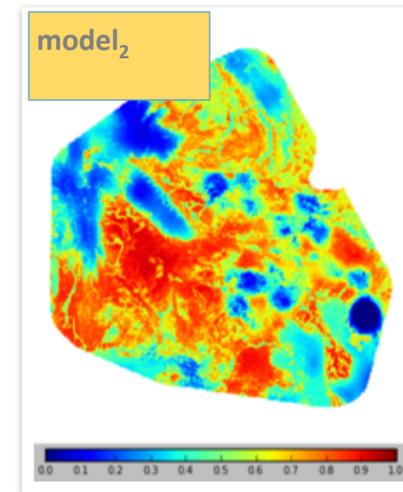
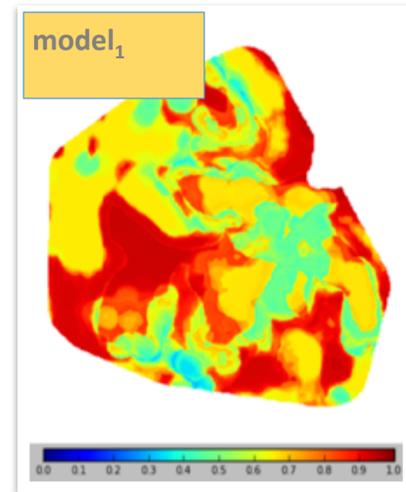


$$f : \mathbb{R}^d \rightarrow [0, 1]$$

Model Inter-comparison



→
Predict
Distribution



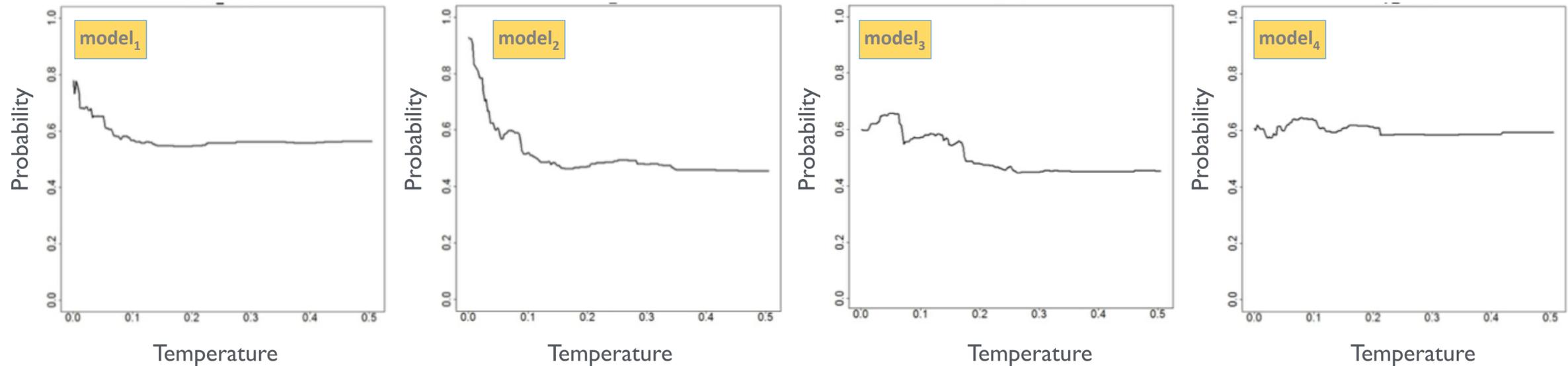
Why are they different?
How do we compare them?

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

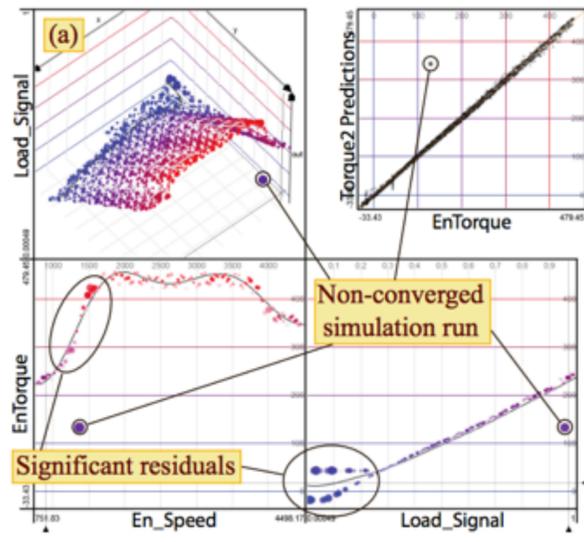
Goals

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

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

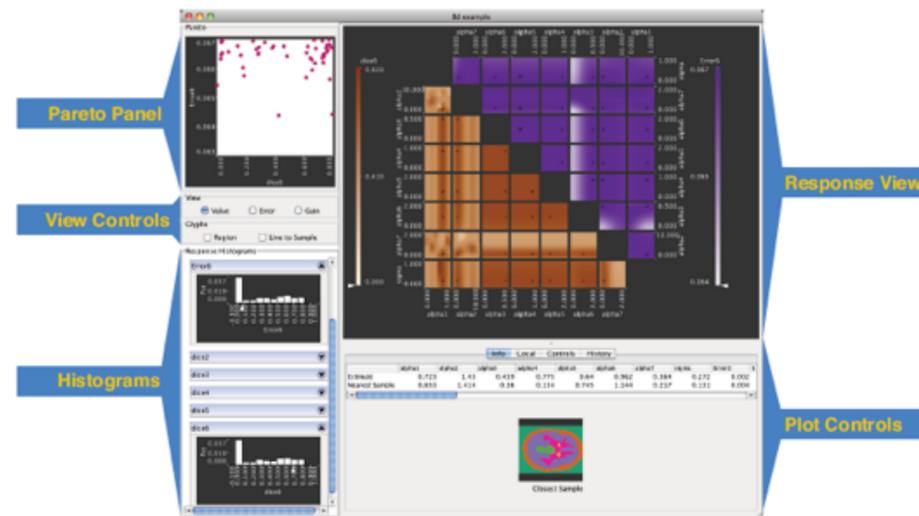
Challenges: Exploring High Dimensional Functions

[HyperMoVal]



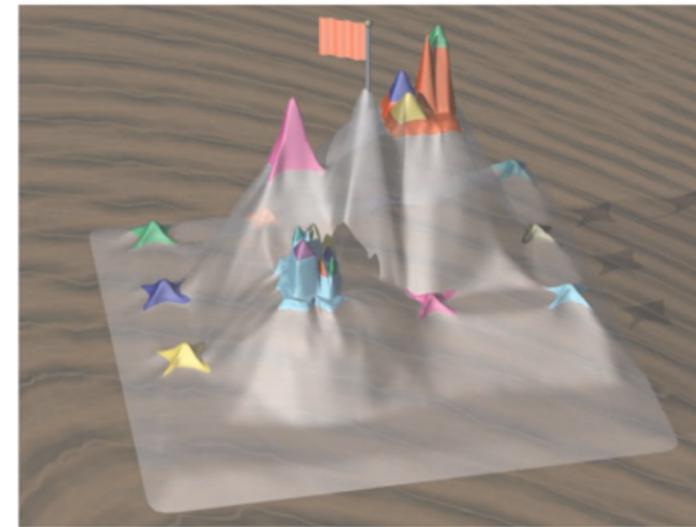
[Piringer et al. 2010]

[Tuner]



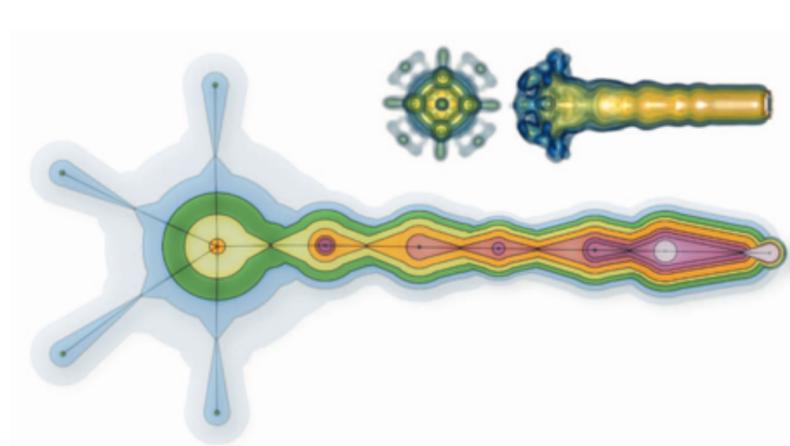
[Torsney-Weir et al. 2011]

[Topological Landscape]



[Weber et al. 2007]

[Topological Spines]



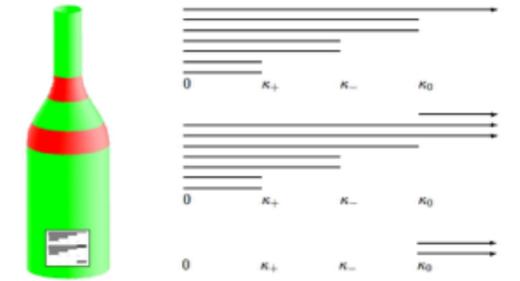
[Correa et al. 2011]

- 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]

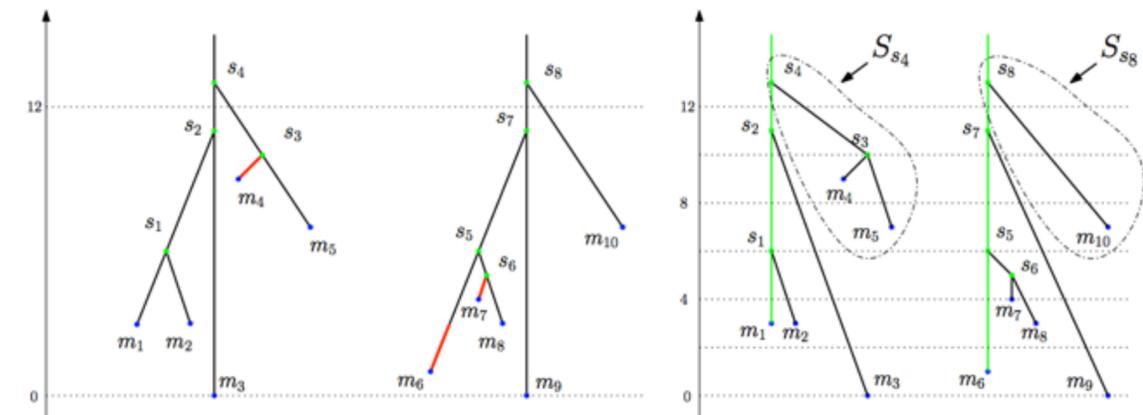
- Distance between topological data structures

- ▶ Merge trees / Reeb graphs

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

- ▶ Extremum graphs

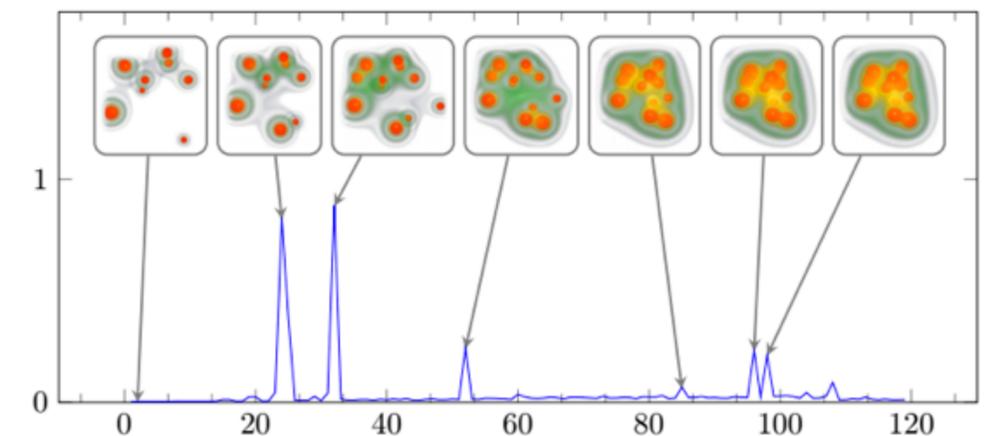
- [Narayanan et al. 2015]



[Beketayev et al. 2014]

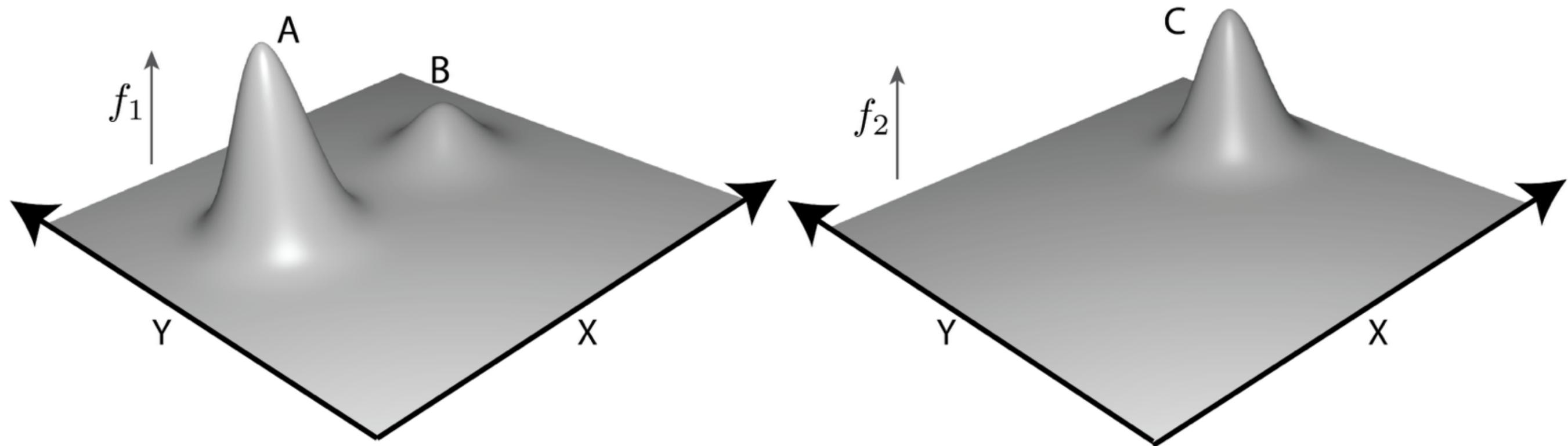
- Shortcomings:

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



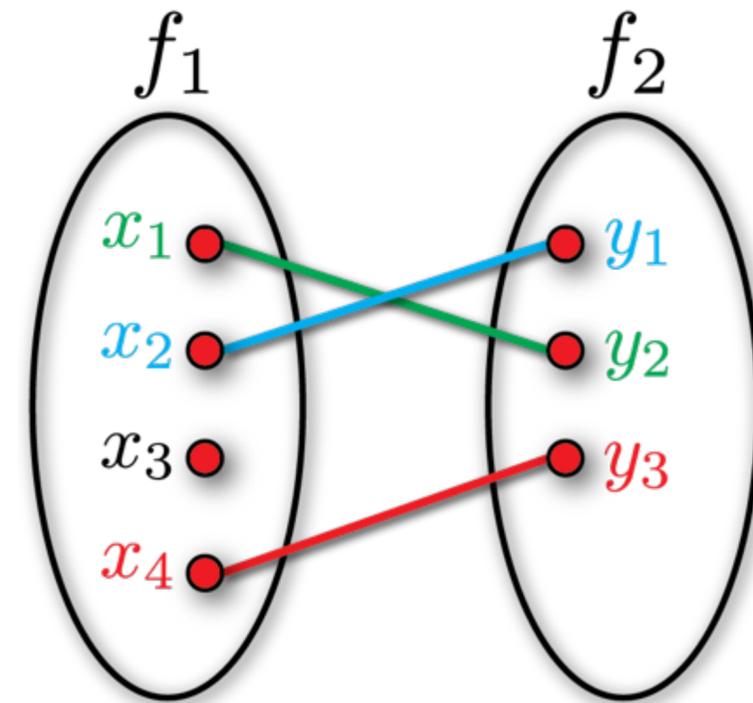
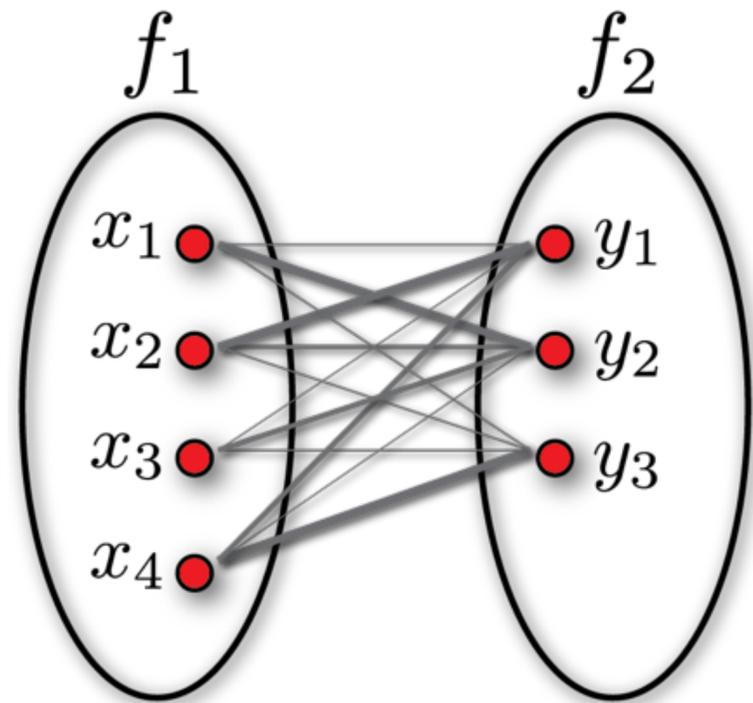
[Narayanan et al. 2015]

Motivation



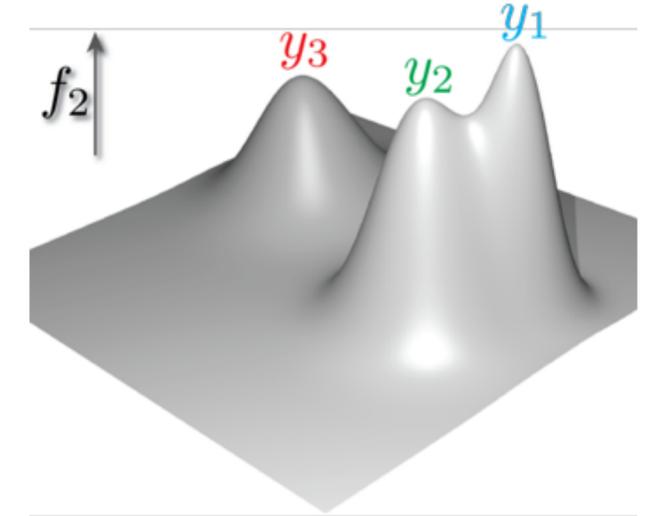
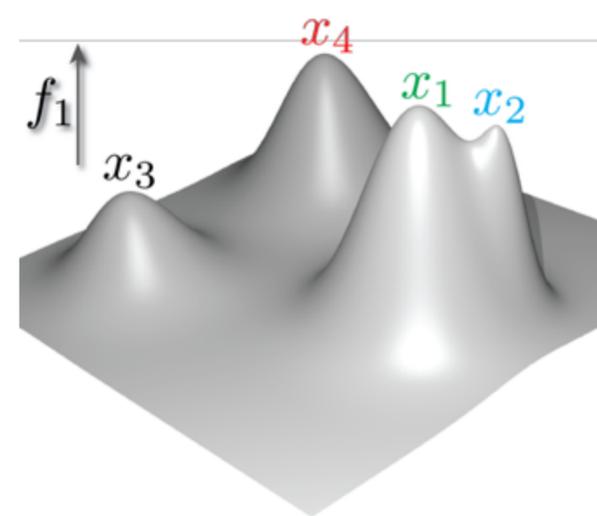
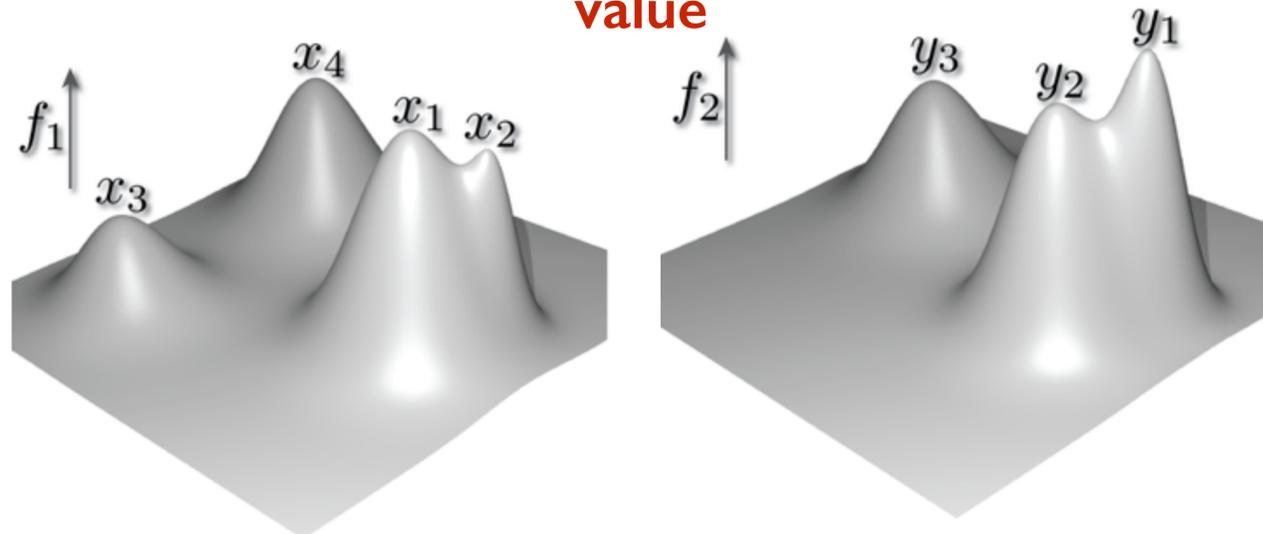
Is there a way to compute a **locality-aware** similarity measure?

Maximum Topology Matching



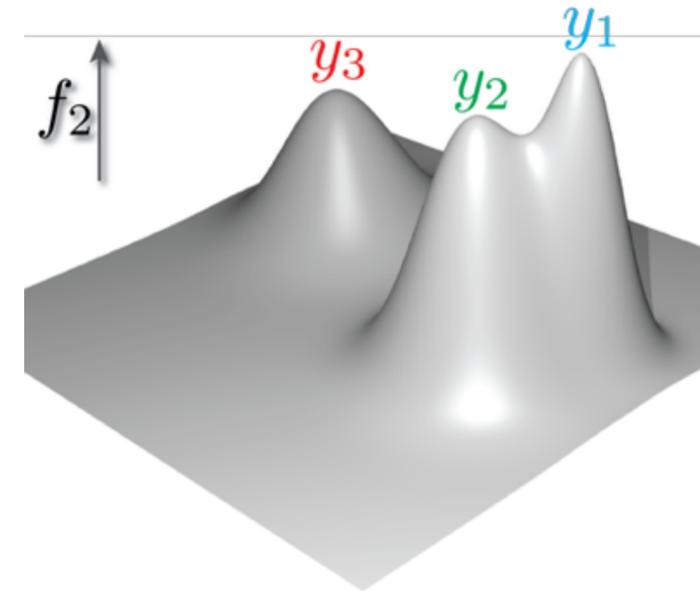
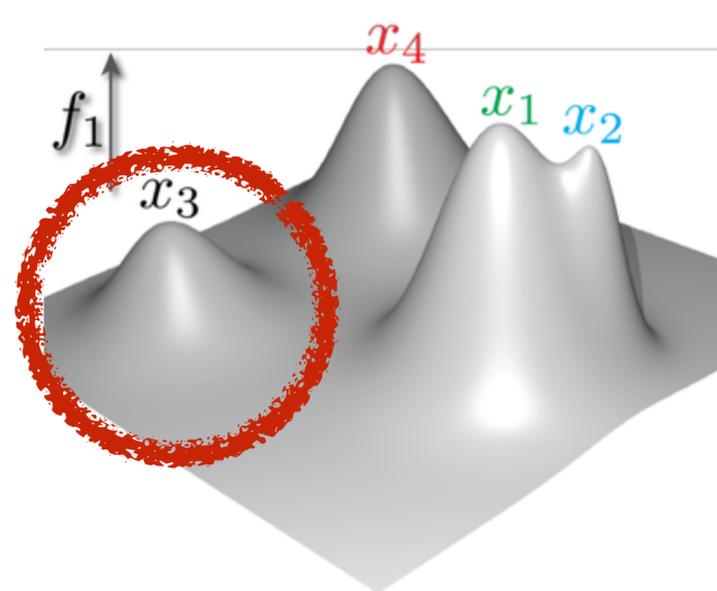
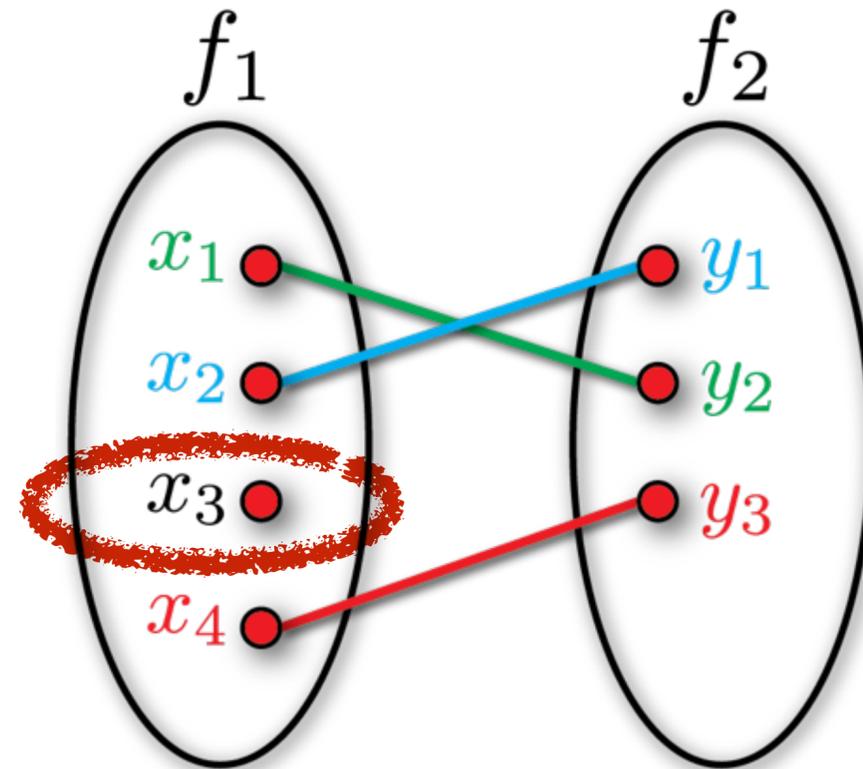
$$w_{a,b} = (1 - \delta_{a,b}) e^{\frac{d_g(a,b)}{r^2}}$$

function value **locality**



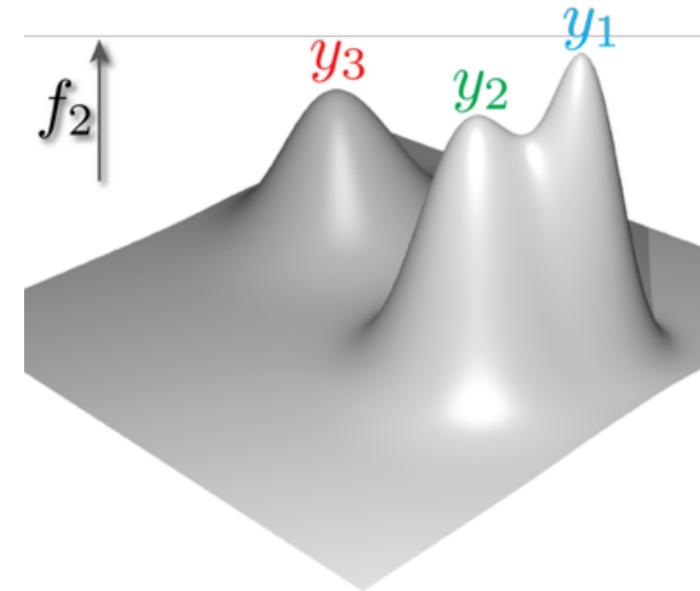
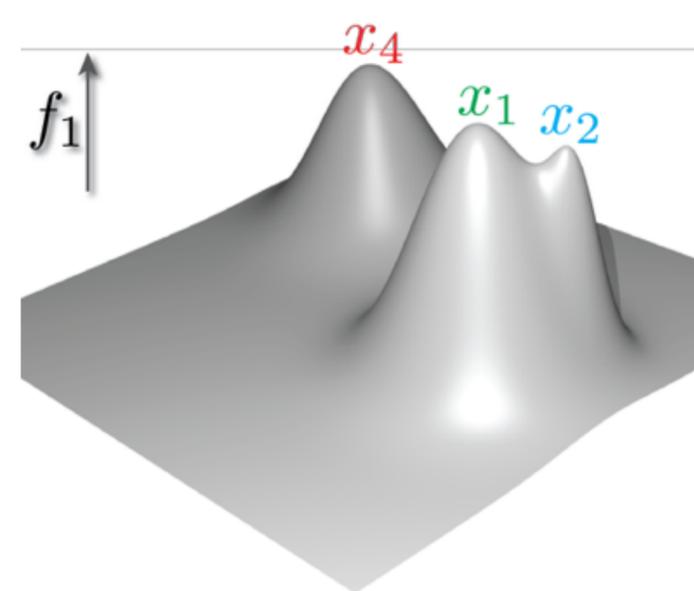
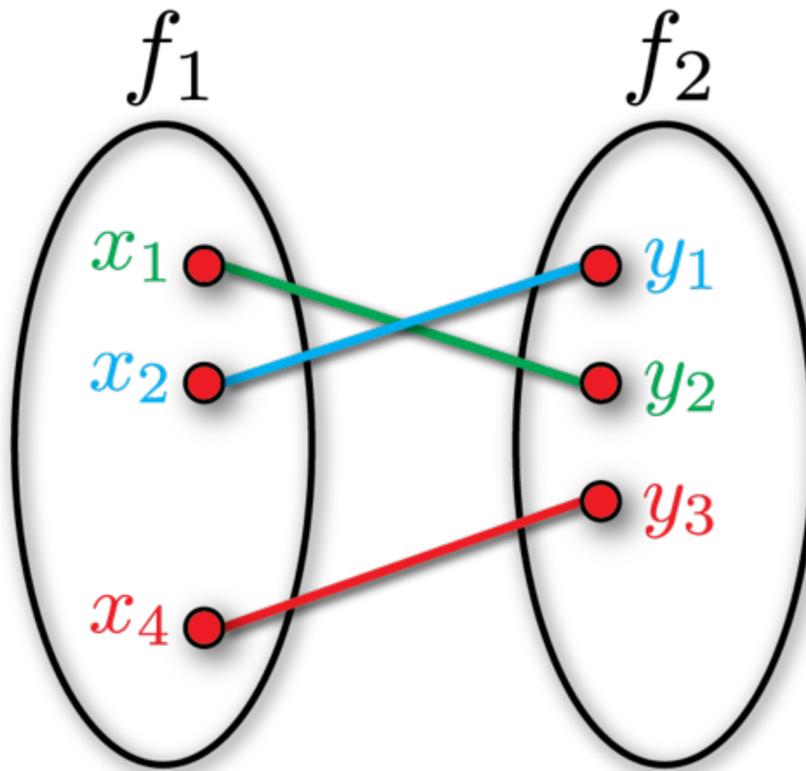
Topological Similarity

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



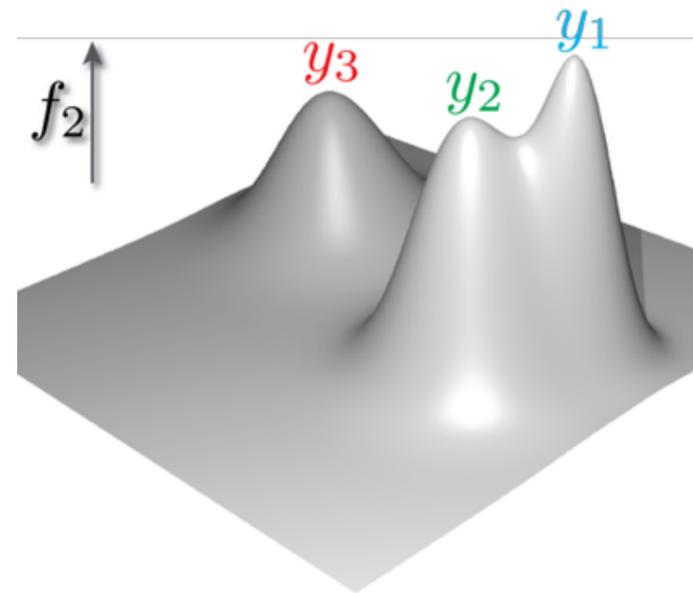
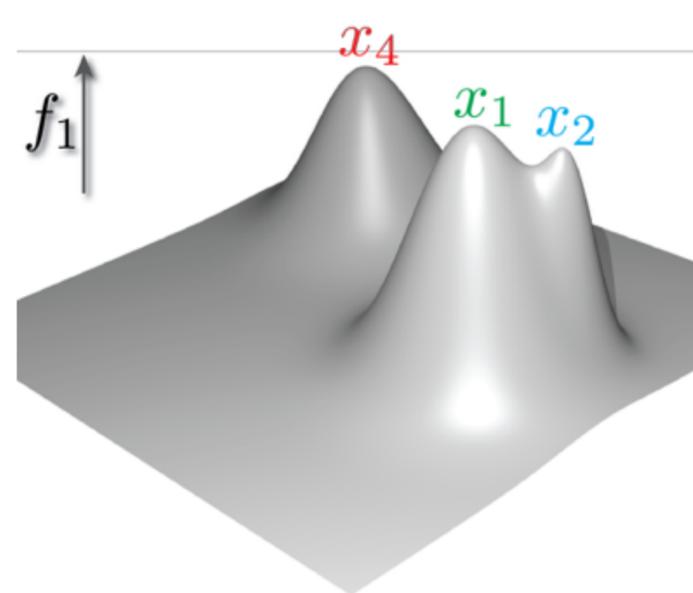
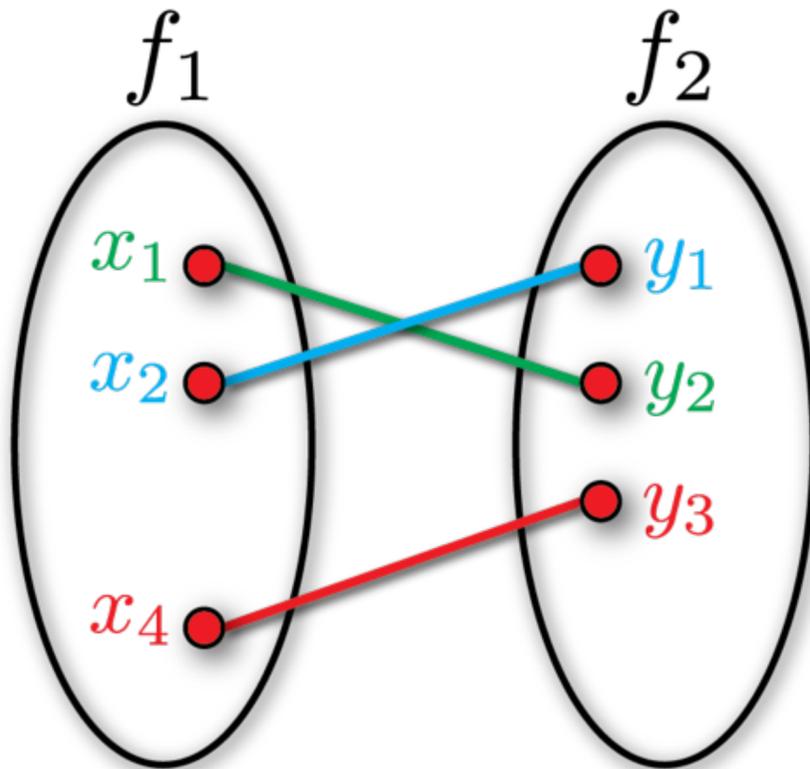
Topological Similarity

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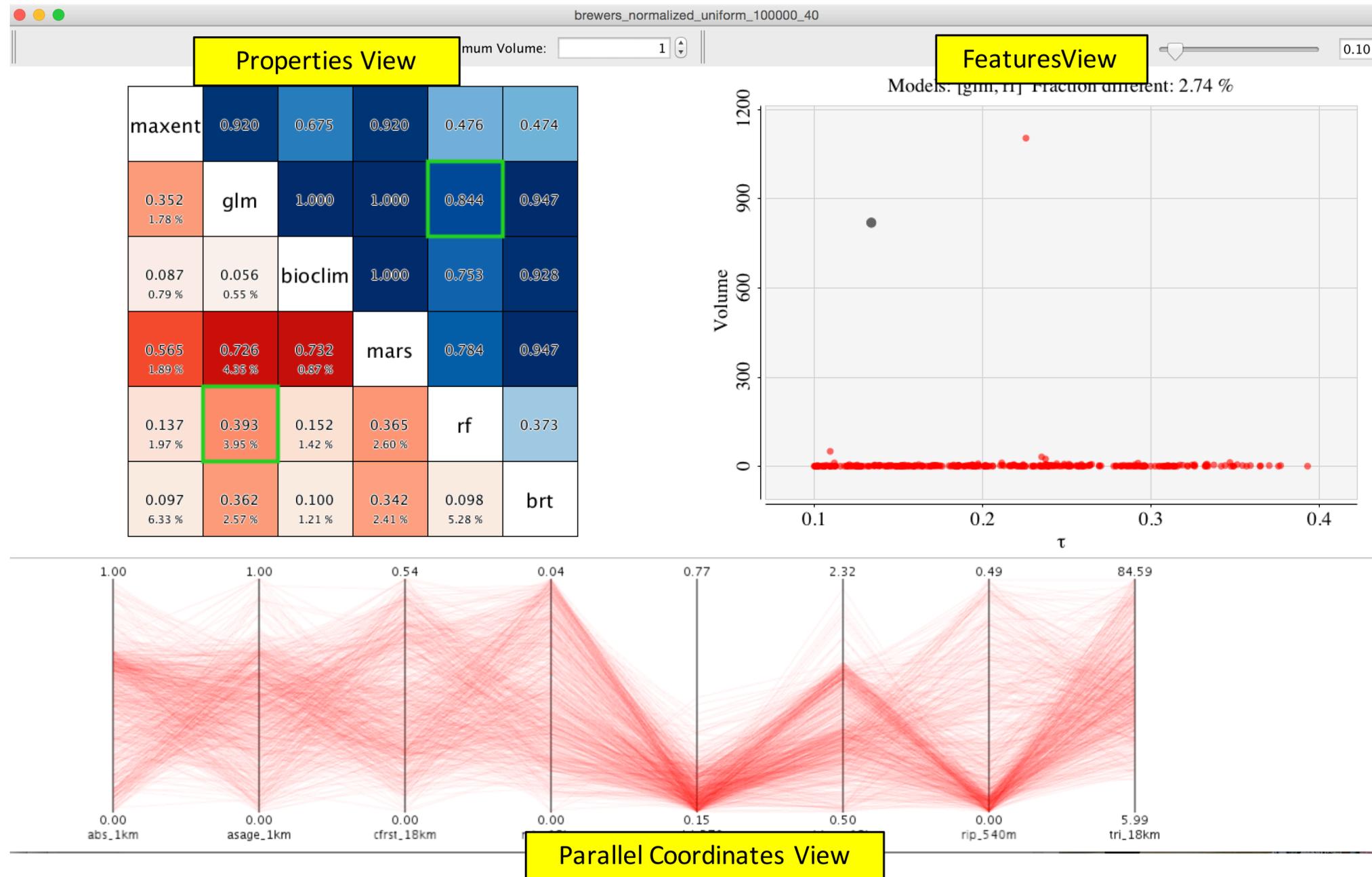
Functional Similarity

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

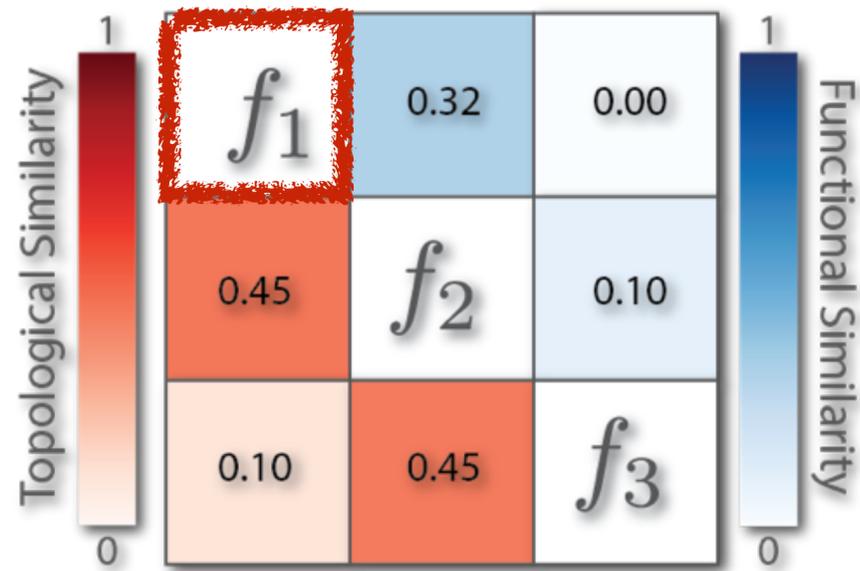
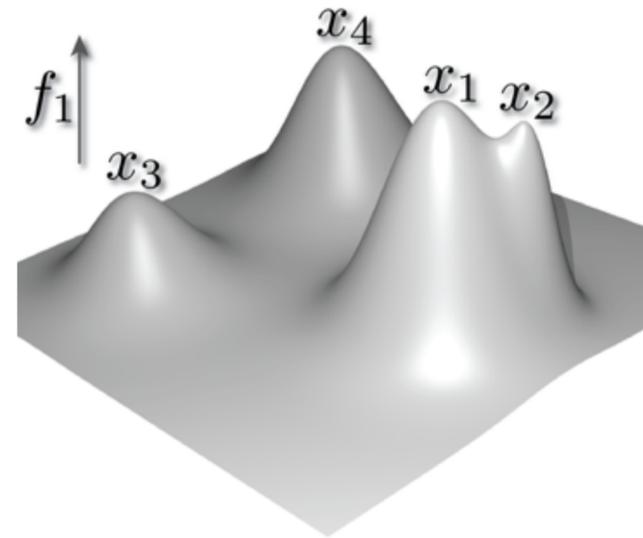


Exploration Framework

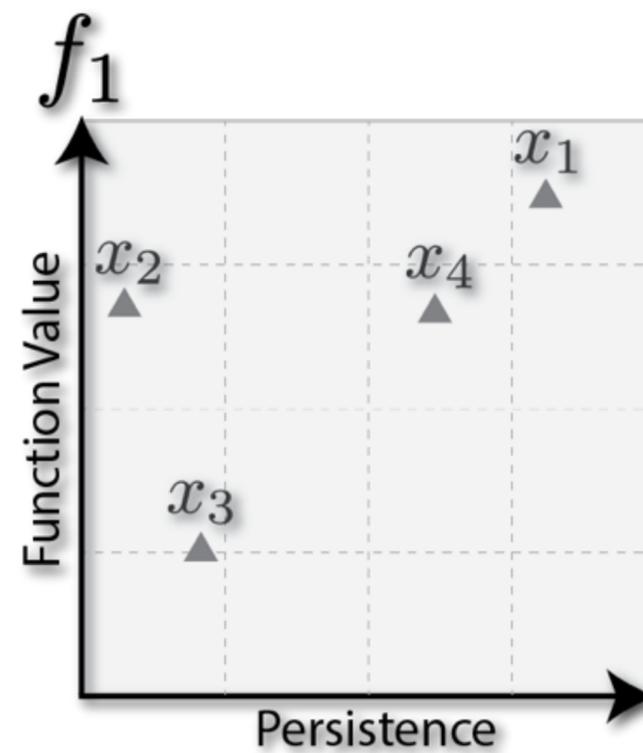
Exploration Framework



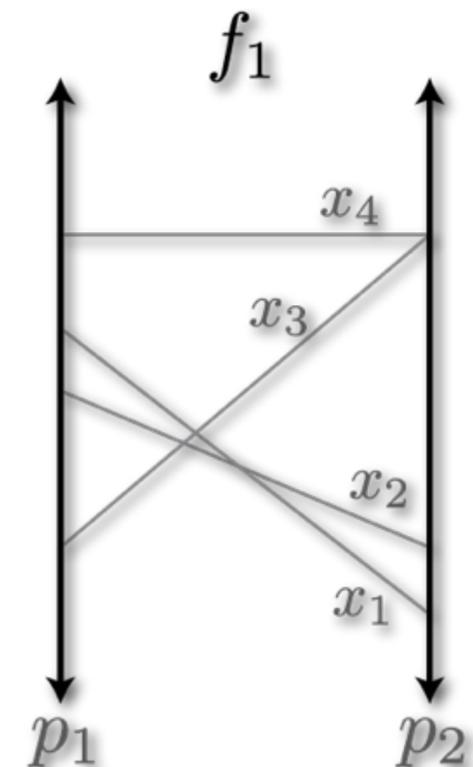
Exploring a SDM



Properties View

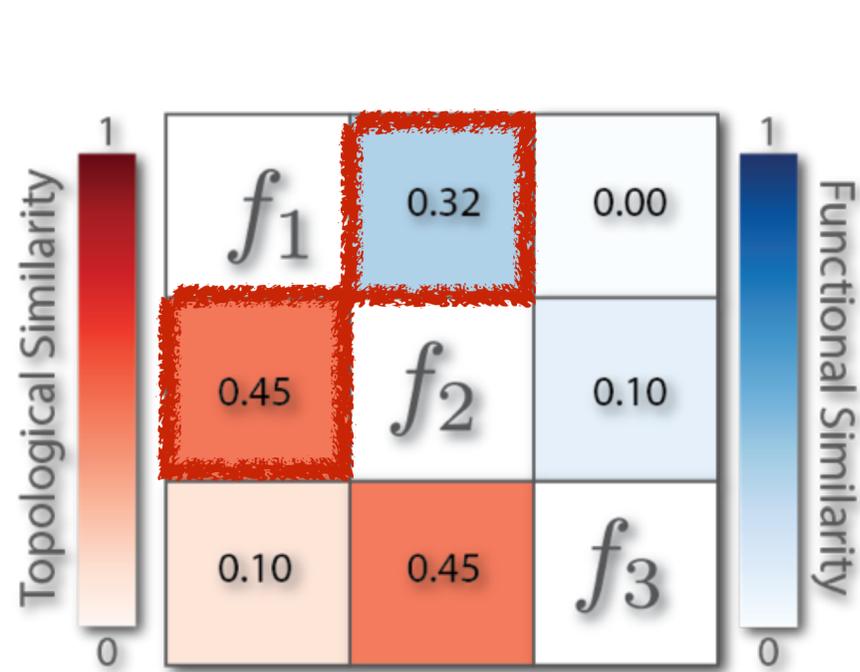
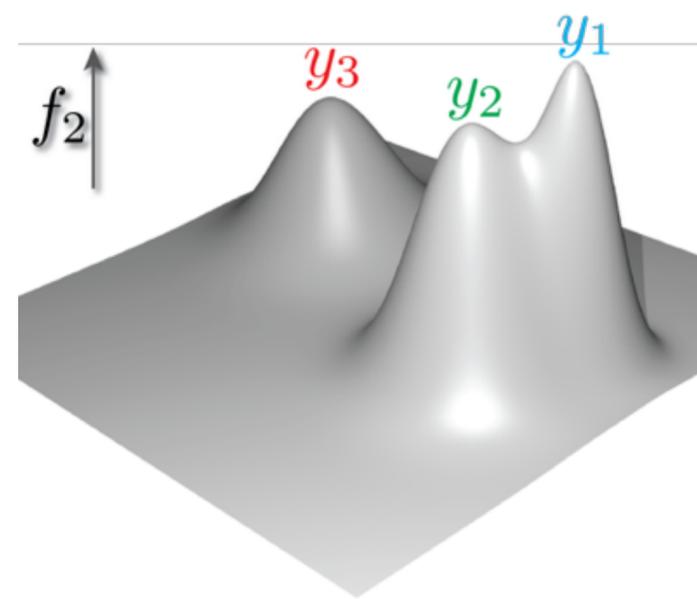
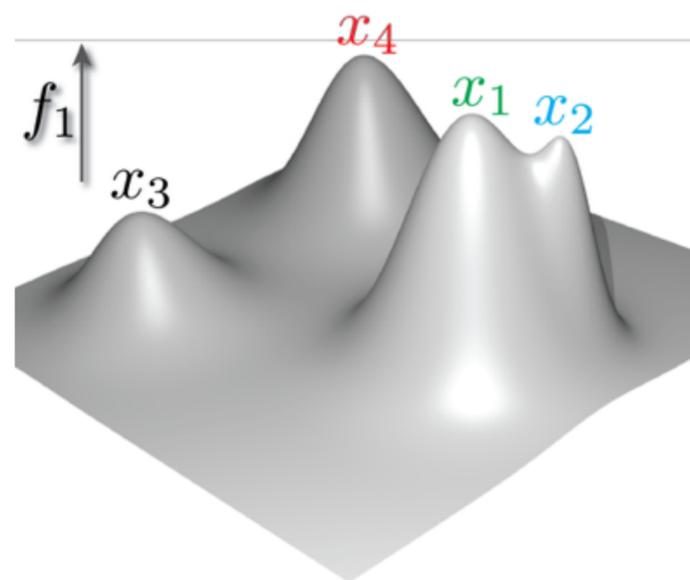


Features View

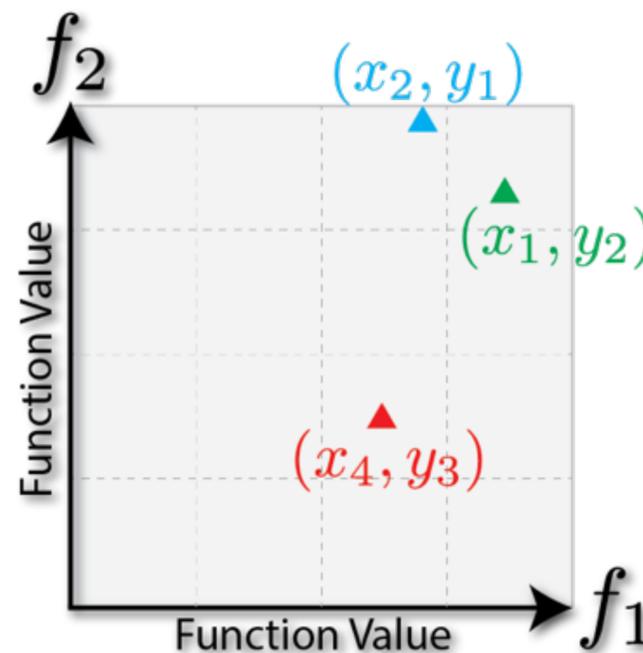


Parallel Coordinates View

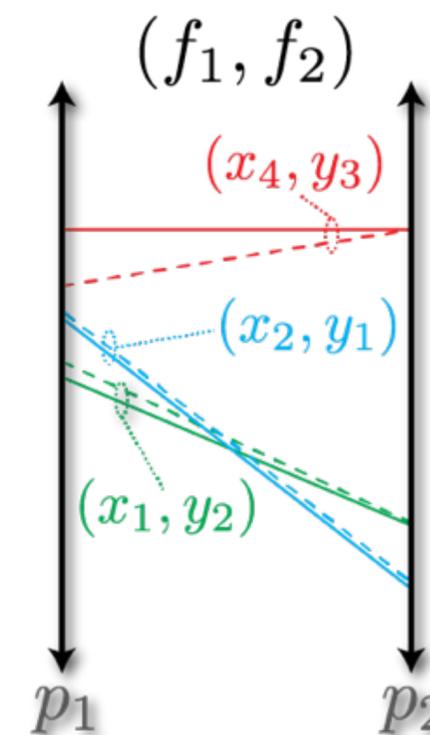
Similarities Between Models



Properties View

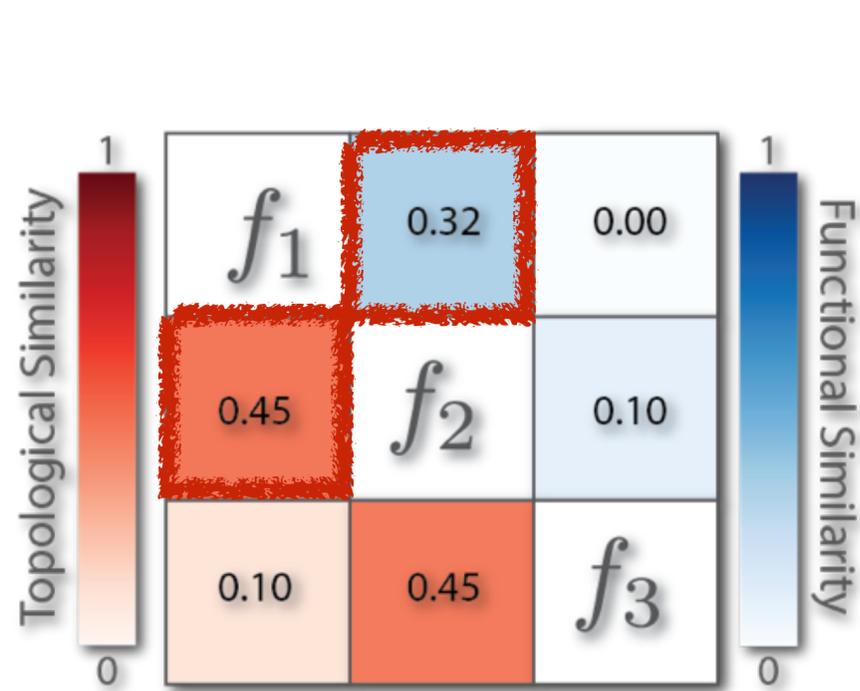
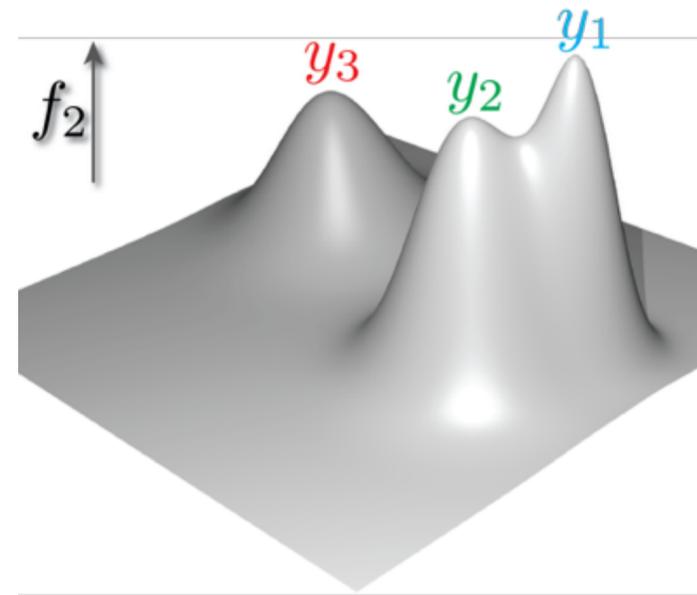
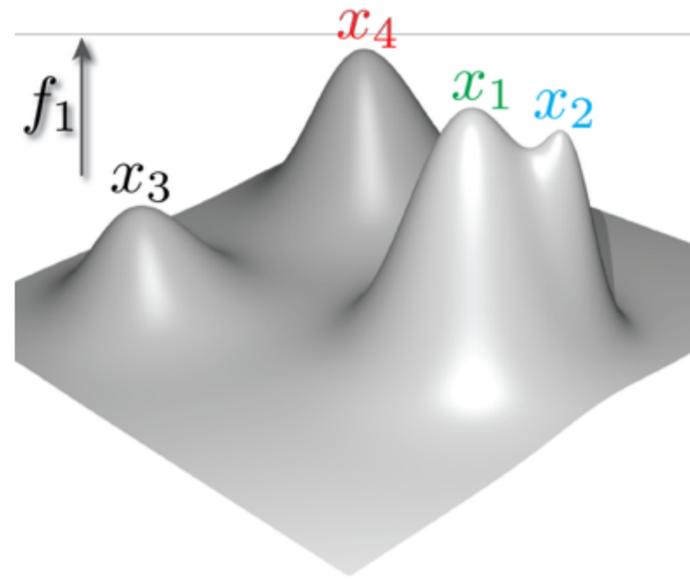


Features View

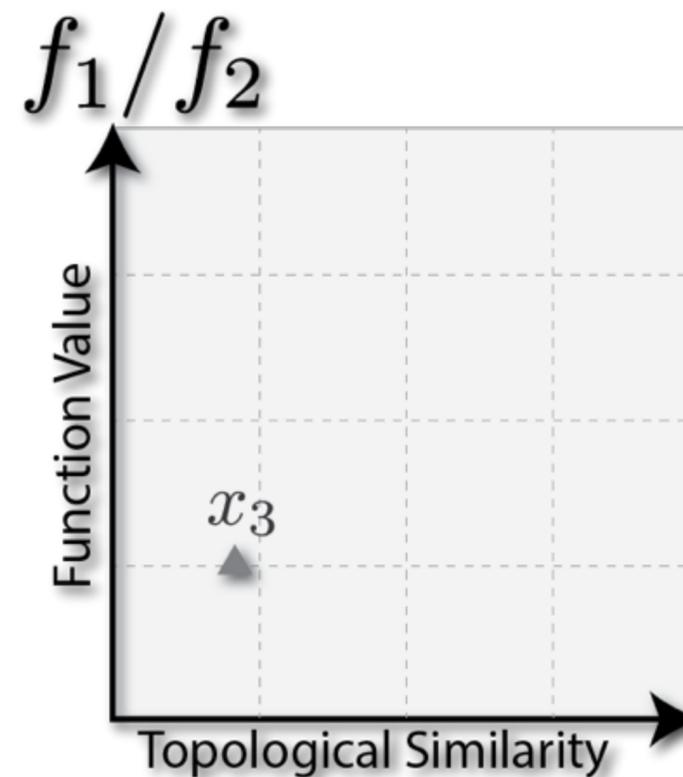


Parallel Coordinates View

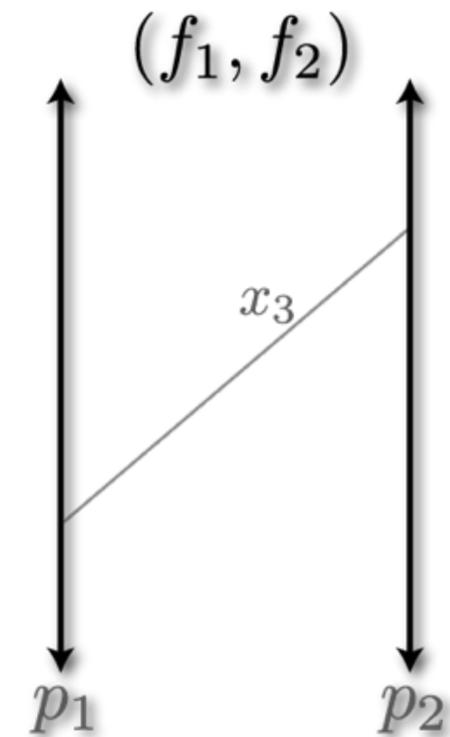
Differences Between Models



Properties View



Features View



Parallel Coordinates View

Case Studies

Data Sets

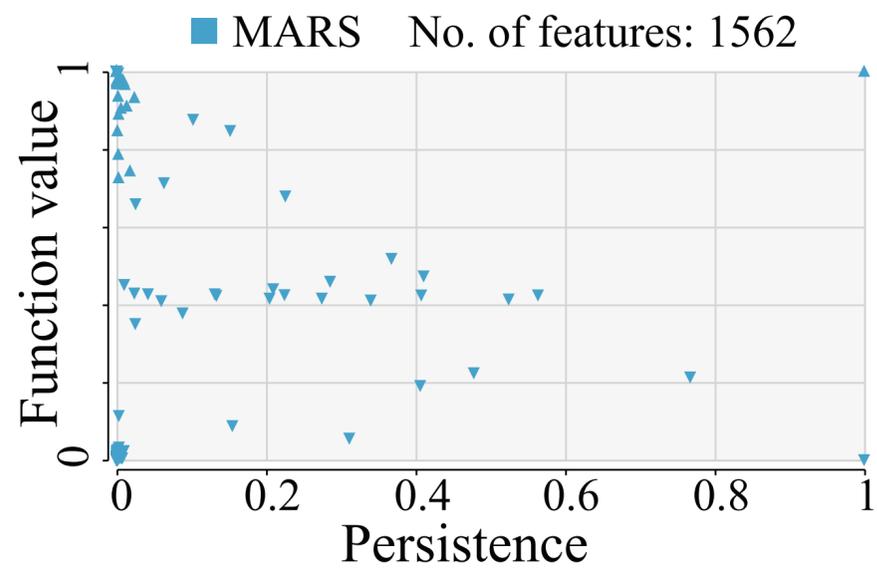
- Brewer's Sparrow
 - 8 predictors
 - 5 models
- Sagebrush
 - 8 predictors
 - 5 models
- Spruce Fir
 - 9 predictors
 - 5 models



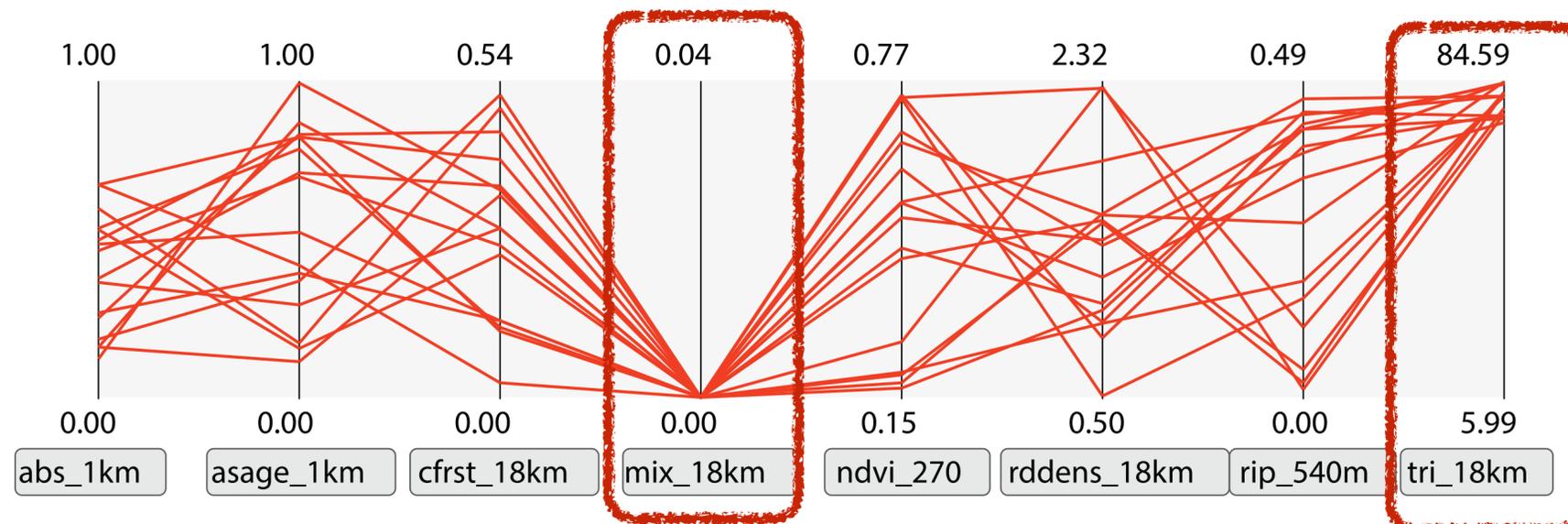
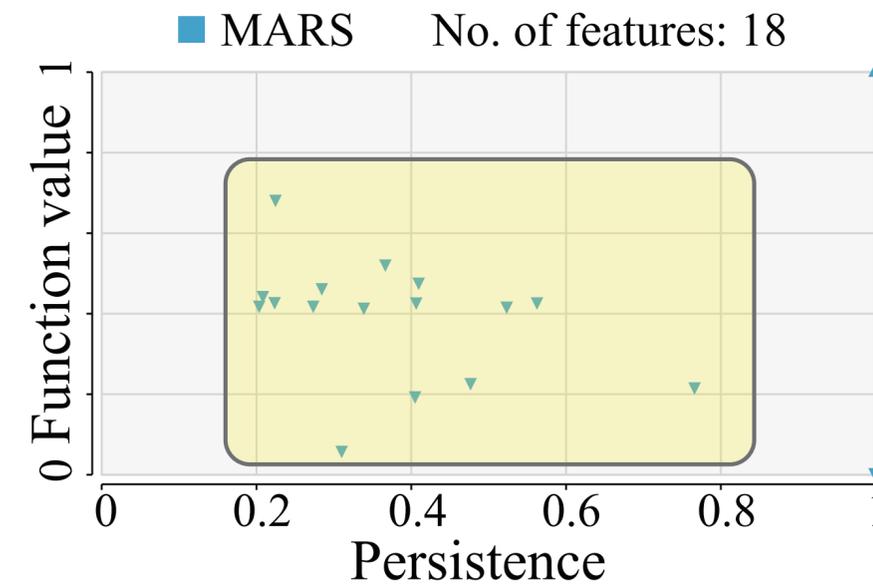
Implementation

- 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 $r = 0.1$

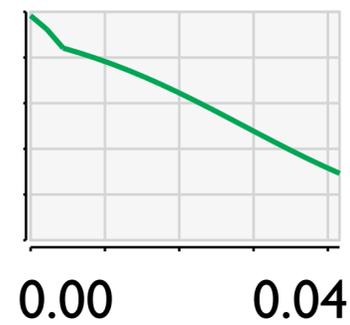
Exploring a *SDM*



➔
Simplification

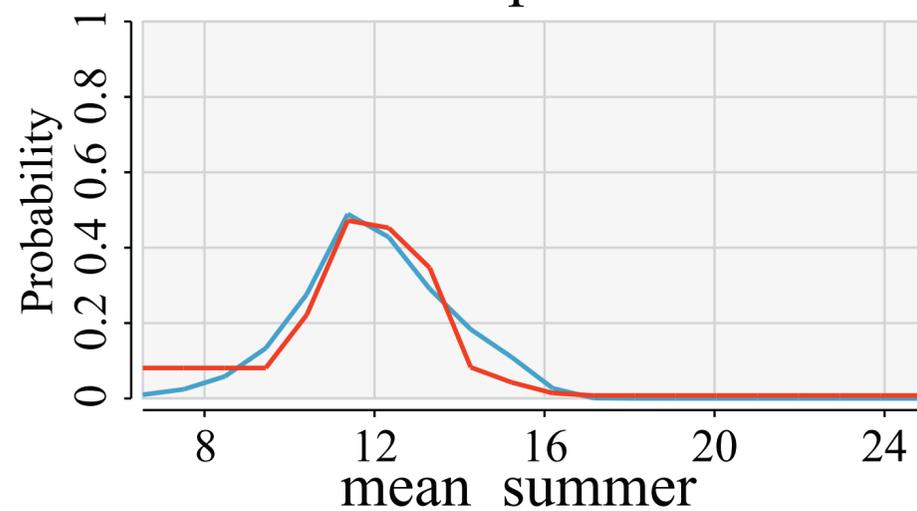


default response curve ➔

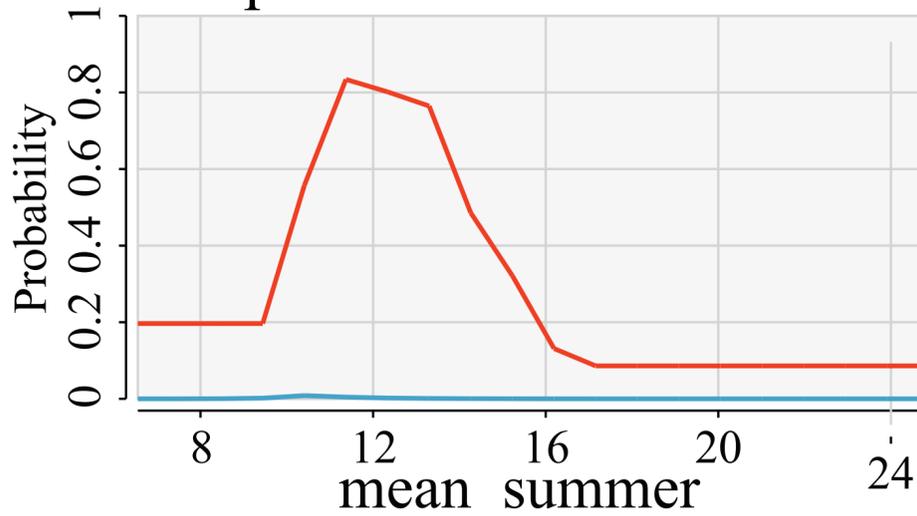


Exploring Differences: MARS vs BRT

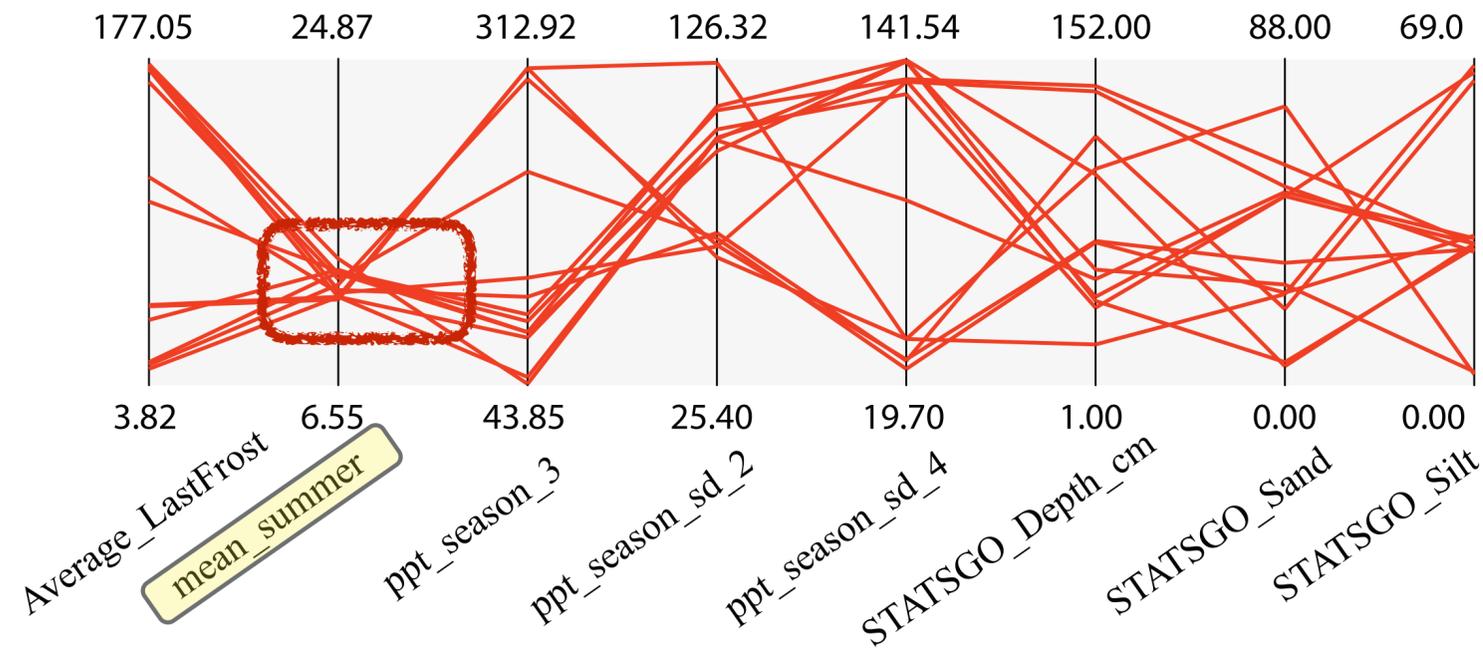
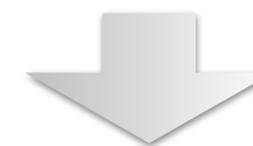
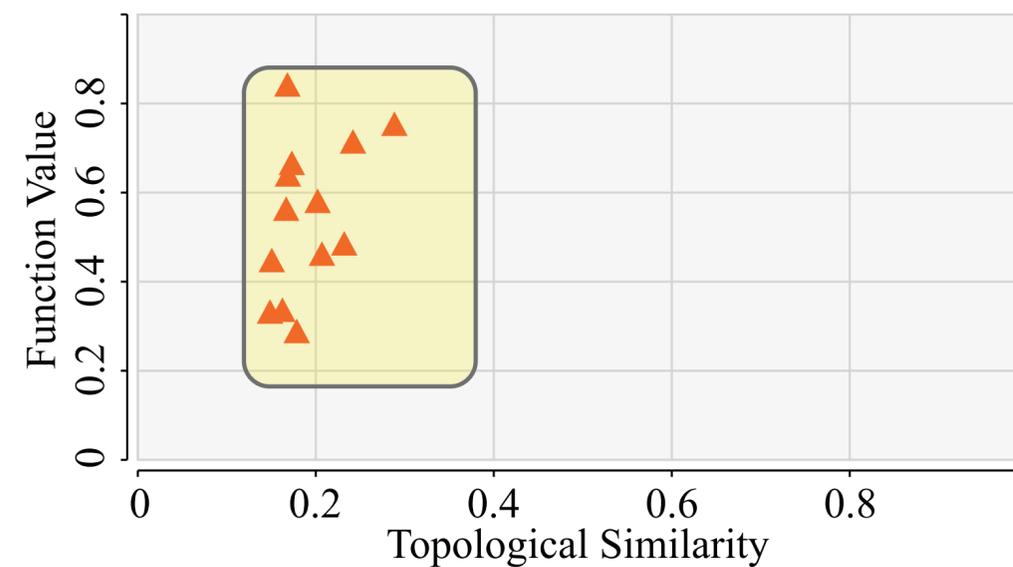
Default Response Curve



Response Curve for Maxima



MARS-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

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- AT&T
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- DOE
- NASA Biodiversity Program award NNH11AS091

Using Maximum Topology to Explore Differences in Species Distribution Models

More Information:

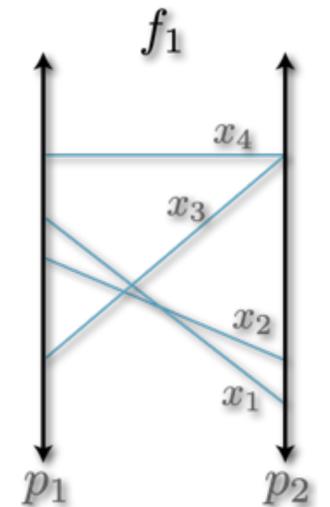
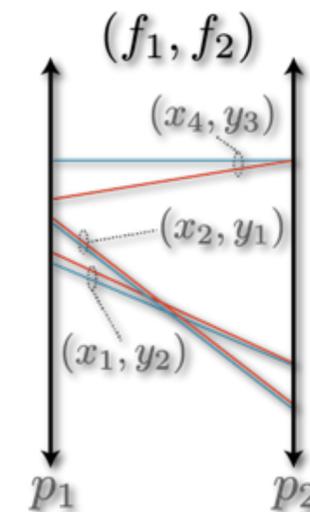
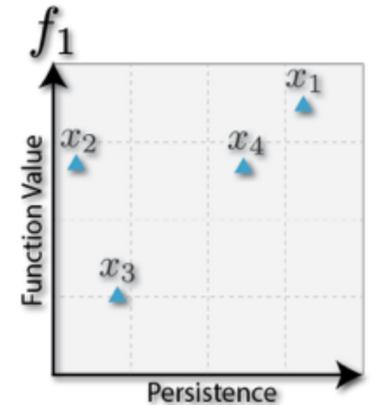
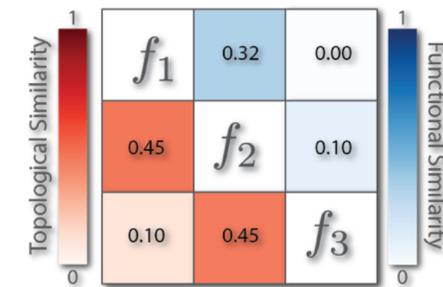
Jorge Poco

jpocom@uw.edu

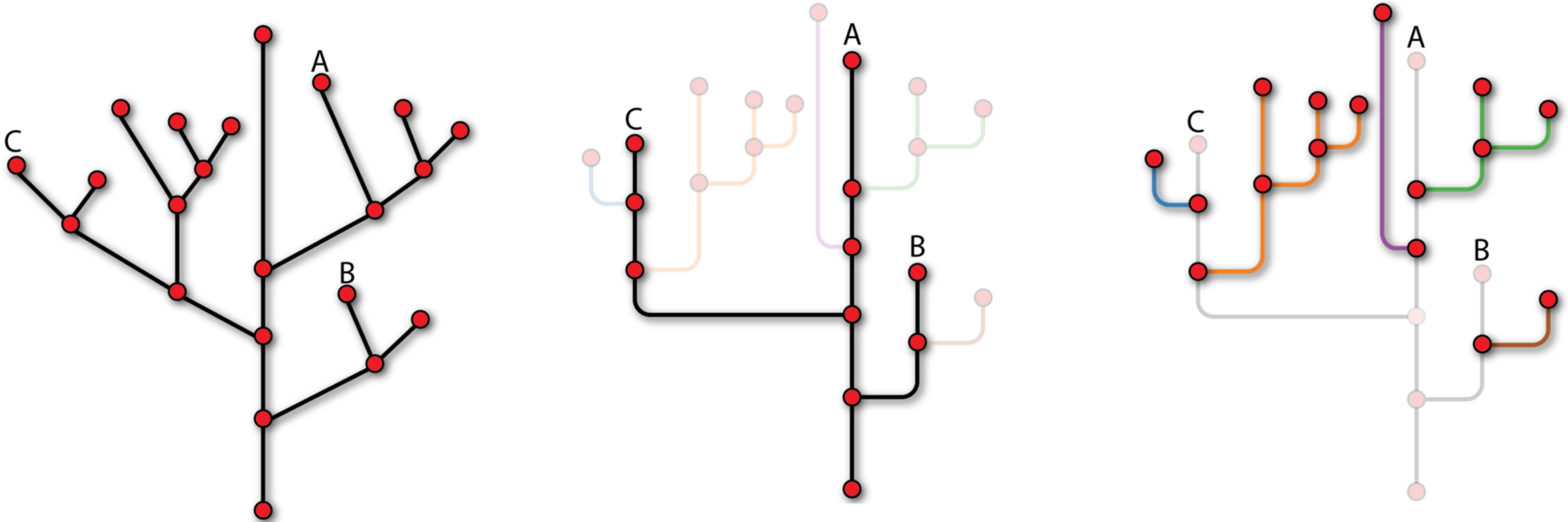
Harish Doraiswamy

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