Structural Analysis of Multivariate Point Clouds using Simplicial Chains

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SPLOM



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■ Use e.g. SPLOM to make sense of data.



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Beyond comparing projection scatterplots?

Typical goal Quantify changes & differences



Our approach

- Combine geometrical (*distance-preserving*) and topological (*structure-preserving*) methods.
- Visualize internal connectivity of a data set as a graph

Topology



- Describe data sets by high-dimensional "holes".
- \blacksquare Hole \approx inhomogeneous region in the data.
- Detection using *persistent homology*.
- Description using *simplicial chains*

Without geometrical information:



What we want

Without geometrical information:



What we want

What we get

Without geometrical information:



What we want

What we get

With geometrical information:



What we want

Without geometrical information:



With geometrical information:



Finding loops



Localization of loops



- Obtain unlocalized loops by traversal.
- Solve all-pairs shortest path problem.
- Obtain localized loops by traversal.

The simplicial chain graph Idealized example



The simplicial chain graph In practice



Data set: Voting data

Alice	+1	+1	 +1
Bob	-1	-1	 —1
Carol	+1	-1	 0
Dave	0	0	 0

- Voting data from the United States House of Representatives.
- About 430 points of dimensions 600–900.
- Data sets from 1990–2011.

Data set: Voting data



Boundaries between parties are defined by "dissenters" from the party line.

Data set: TAO

Tropical atmosphere ocean project



- El Niño phenomenon.
- Continuous data stream from all buoy moorings.
- 5-dimensional feature space: Wind velocities, humidity, air temperature, and sea surface temperature.

Data set: TAO



1993, 1996: No El Niño phenomenon in data set.
1994–1995, 1997–1998: El Niño phenomenon.

Lessons learned

- Features obtained via persistent homology are suitable for comparative analysis.
- Visualization of internal point cloud structure quickly becomes too abstract.
- Thus: Aim to quantify differences in persistent homology using well-defined metrics.

Summarize data even further



Application

Quantifying differences in models for multivariate data



Application

A framework for comparing dimensionality reduction methods

- Define data descriptors on data set.
- Calculate their persistent homology.
- Compare their "topological fingerprints".



Conclusion

- Simplicial chain graphs as a visual metaphor for structures in multivariate point clouds.
- Aspects: Comparison and quantifying differences graphically.
- Recent work concentrates on metric quantification using persistent homology.
- Open question: What "interesting" structures in high-dimensional space can we capture?