

Additional techniques

Thomas Torsney-Weir

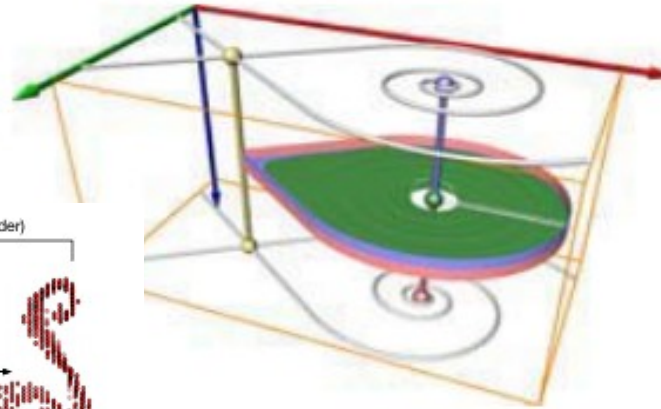
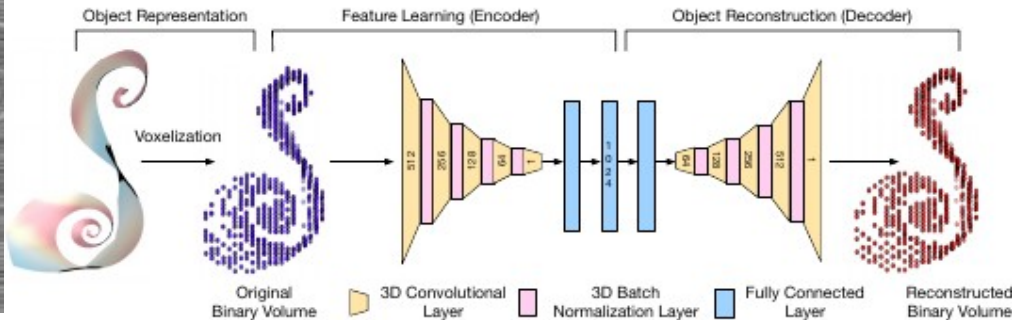
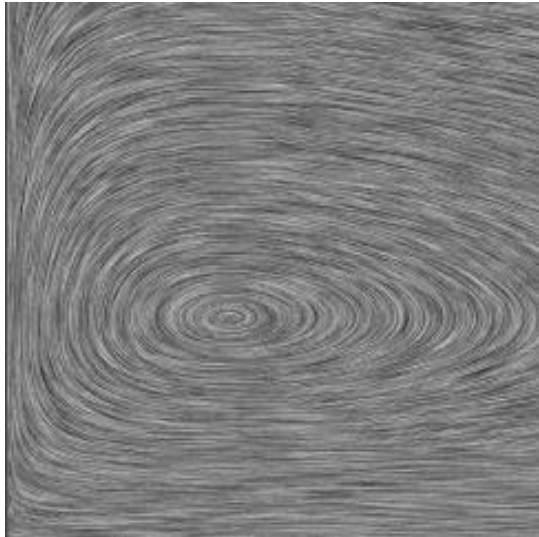
The story so far

- Direct flow visualiation - glyphs
- Indirect flow visualization - streamlines, pathlines, streaklines

What else is there?

Other techniques

- Line integral convolution (LIC) - textures
- Feature-based flow visualization
- Topology-based flow visualization

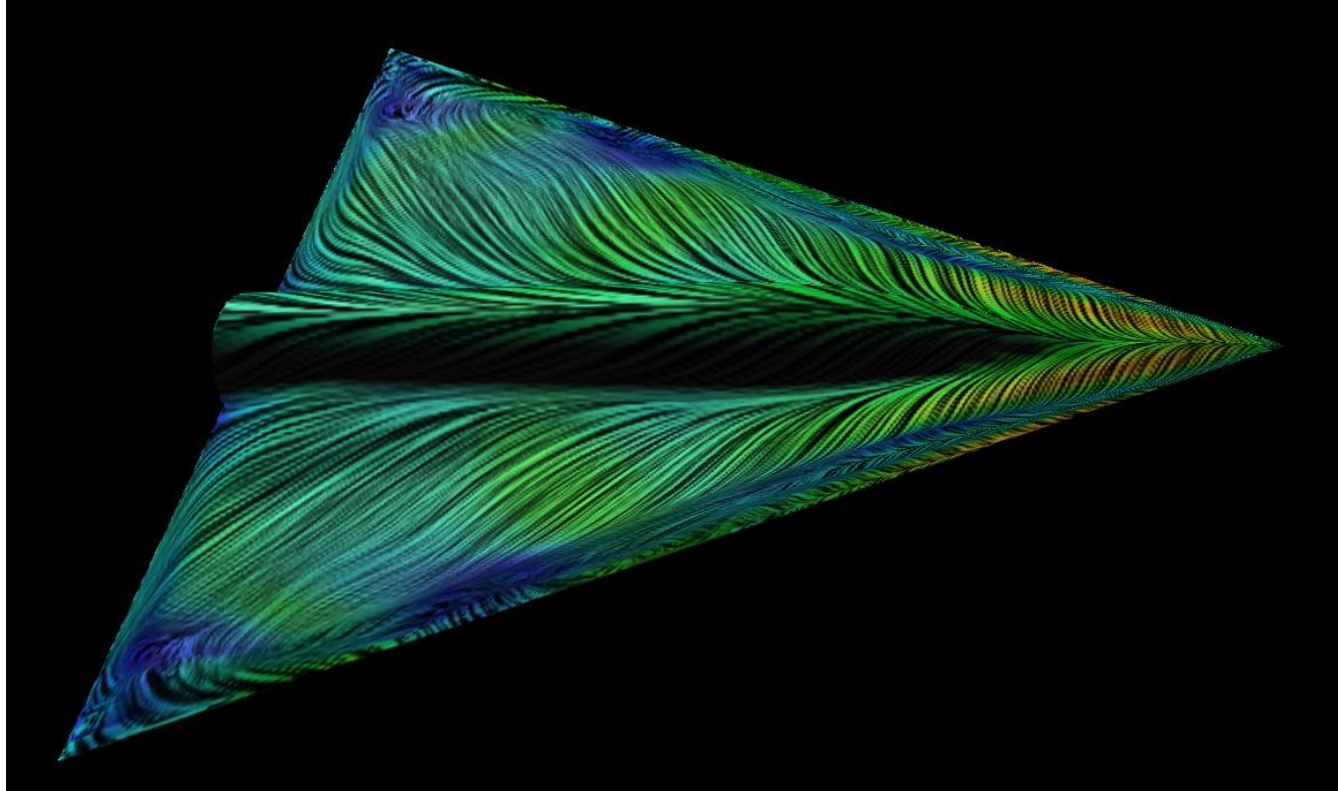


Line integral convolution

Line integral convolution

Useful for getting a general overview of the flow

- Convolution of white noise and a smoothing filter
- Streaks from streamlines through noise



How?

Inputs:

- flow data ($v(x)$)
- white noise ($n(x)$)
- filter ($h(t)$) --- Gaussian is popular choice

Algorithm for output image $lic(x)$:

For each pixel location x :

$s_x(s) \leftarrow$ find streamline through x

$lic(x) \leftarrow 0$

For $u = 0$ to L // length of line

$lic(x) += n(s_x(u)) * h(u)$

$$lic(x) = \int n(s_x(u)) h(u) du$$

Flow Data

Streamline
(DDA)

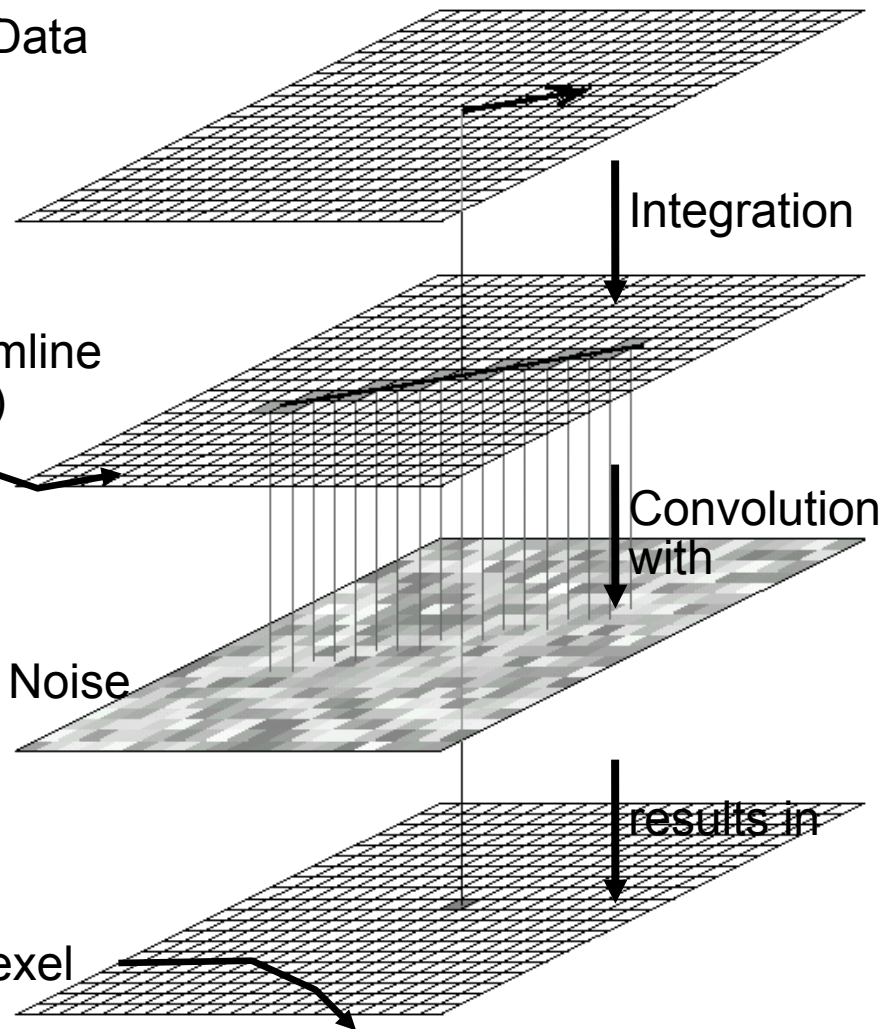
White Noise

LIC Texel

Integration

Convolution
with

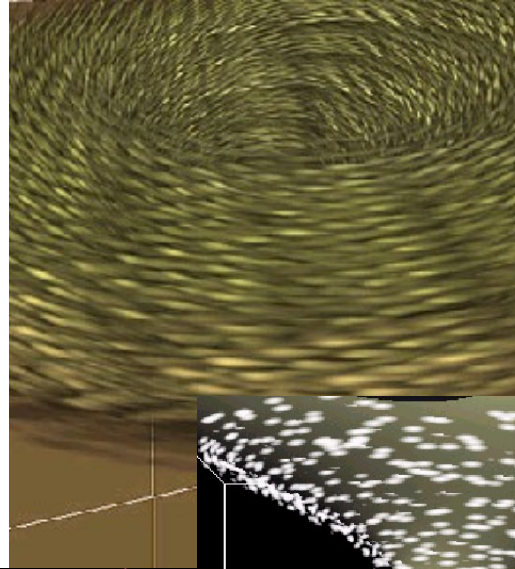
results in



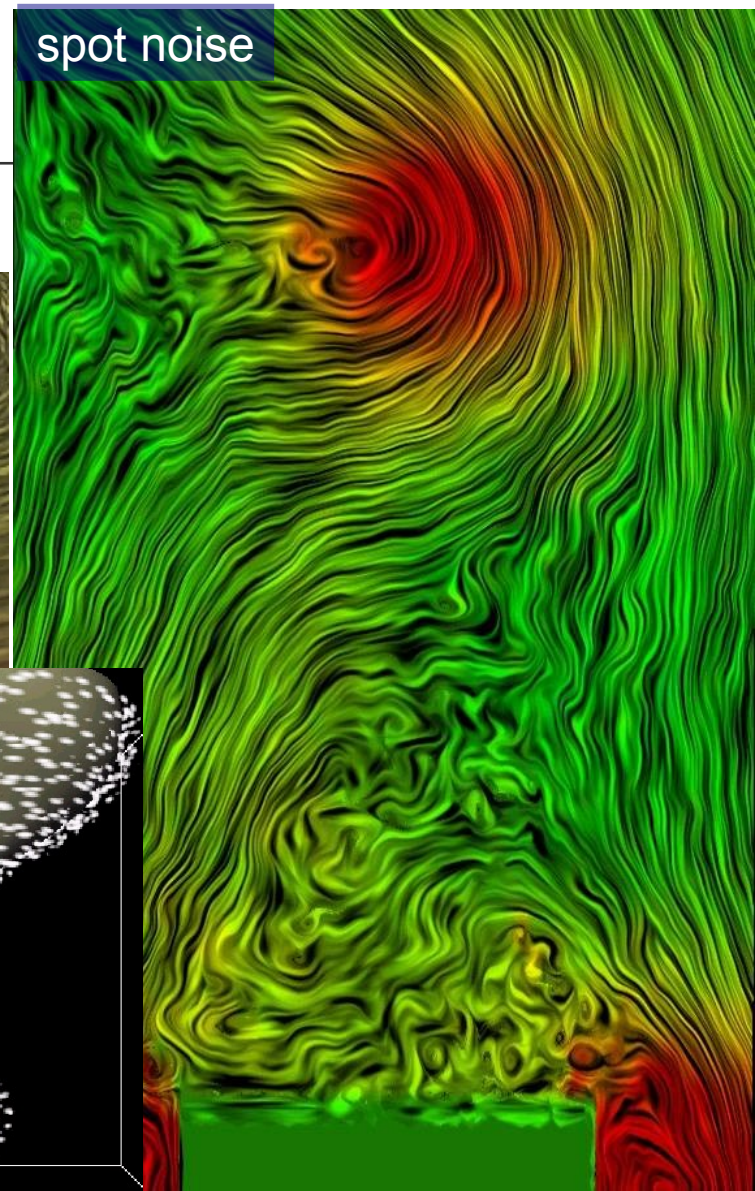
Alternatives to LIC

- Spot noise
- Textured splats
- Particle systems (weather maps)
- Flow volumes

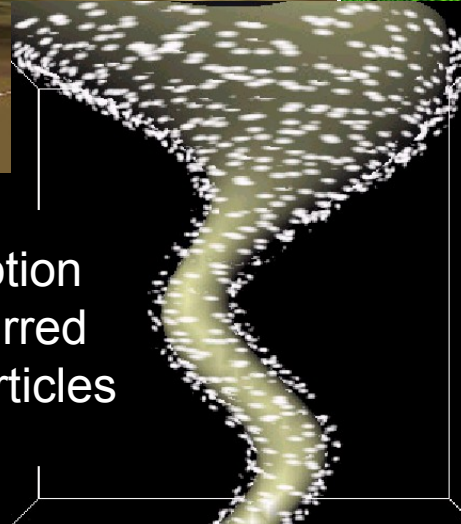
textured splats



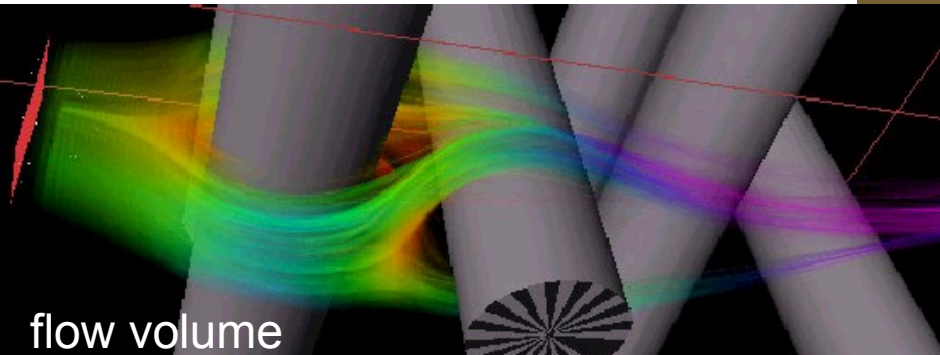
spot noise



motion
blurred
particles



flow volume

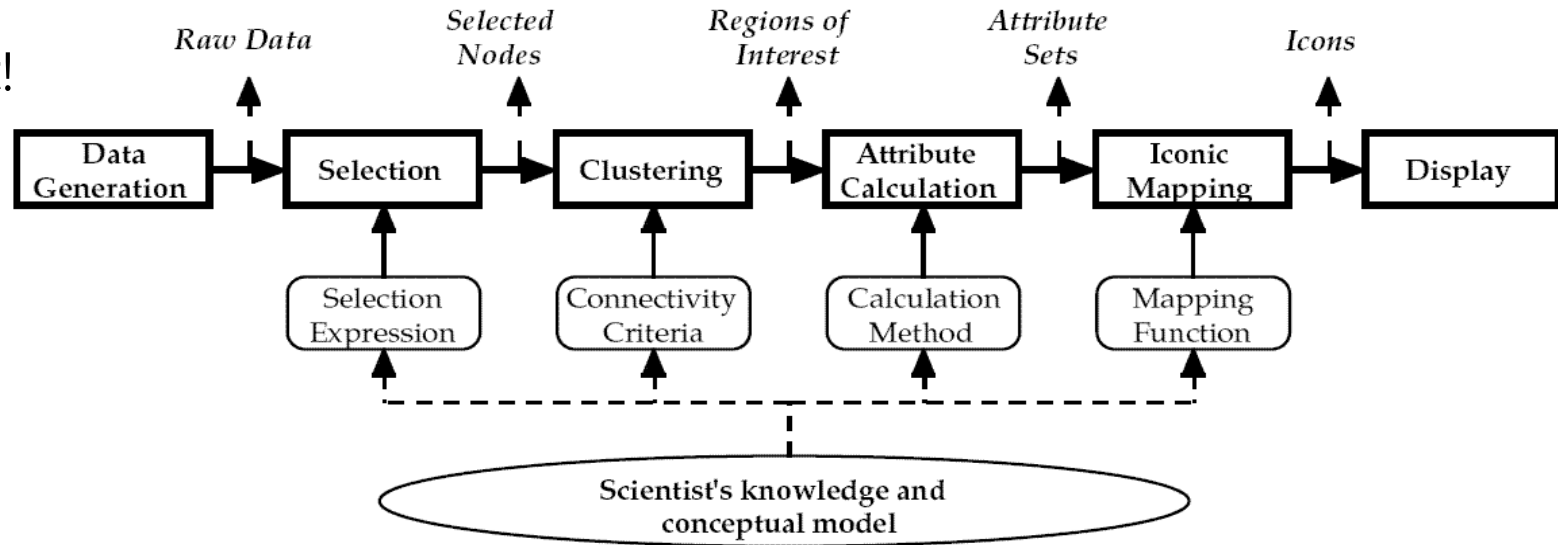


Feature-based flow visualization

Feature-based flow visualization

Feature : a prominent or distinctive aspect, quality, or characteristic of the flow field

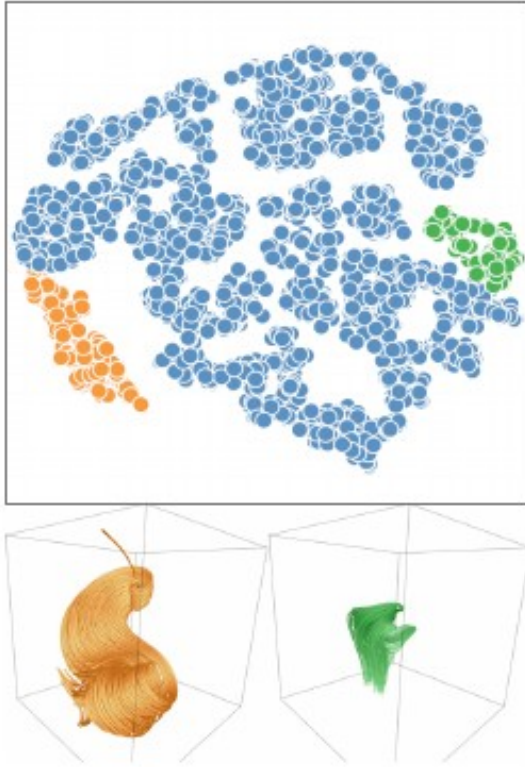
- 1) Identify what are "interesting" features for the user
- 2) Design an algorithm to identify/extract these features
- 3) Show them to the user
- 4) Profit!



Motivation

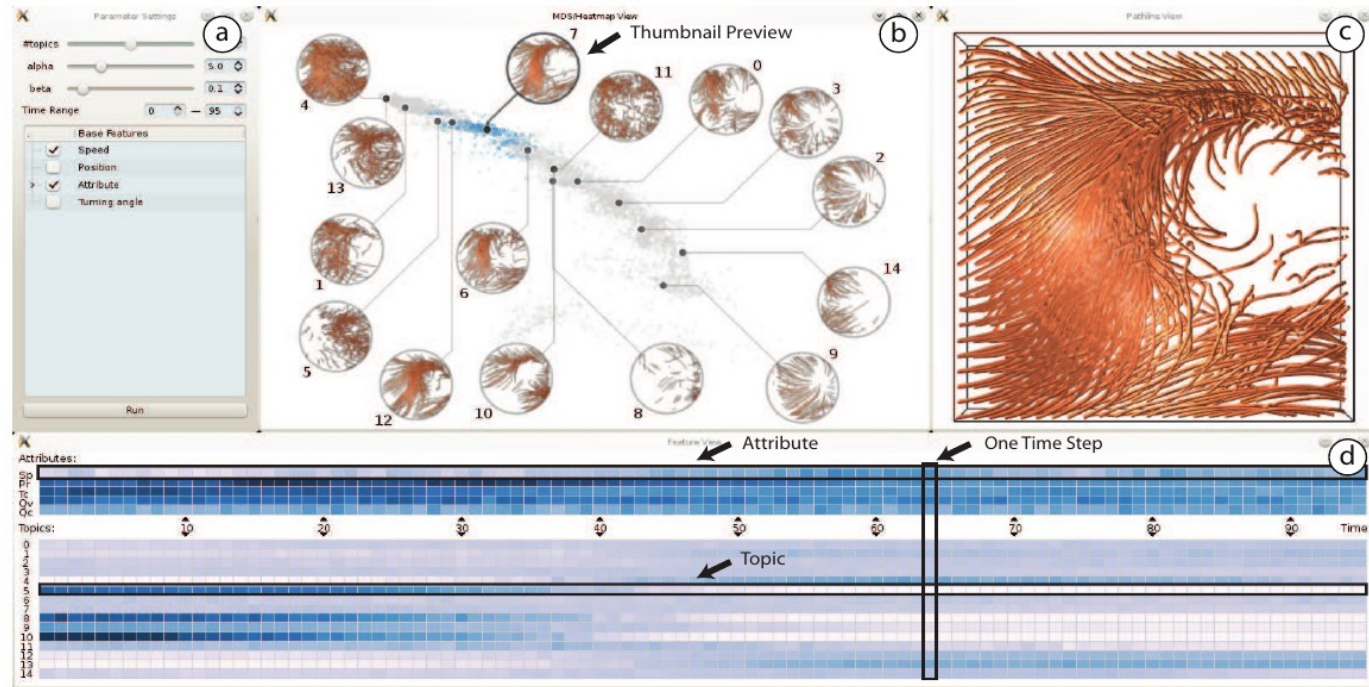
- *data reduction* : original data set is represented with important features
- *perception* : visualization of 3D and 4D flow is problematic in the absence of feature-based techniques
- *new insight* : "new" characteristics of flow can be observed
- *technical advantages* : less memory consumption, faster interaction and rendering
- Good when you have clear idea of what to show

Examples



Han, Jun, Jun Tao, and Chaoli Wang. "FlowNet: A Deep Learning Framework for Clustering and Selection of Streamlines and Stream Surfaces." IEEE Transactions on Visualization and Computer Graphics, 2018, 1-1.

<https://doi.org/10.1109/TVCG.2018.2880207>.



Hong, Fan, Chufan Lai, Hanqi Guo, Enya Shen, Xiaoru Yuan, and Sikun Li. "FLDA: Latent Dirichlet Allocation Based Unsteady Flow Analysis." IEEE Transactions on Visualization and Computer Graphics 20, no. 12 (December 2014): 2545-54.

<https://doi.org/10.1109/TVCG.2014.2346416>.

Topology-based flow visualization

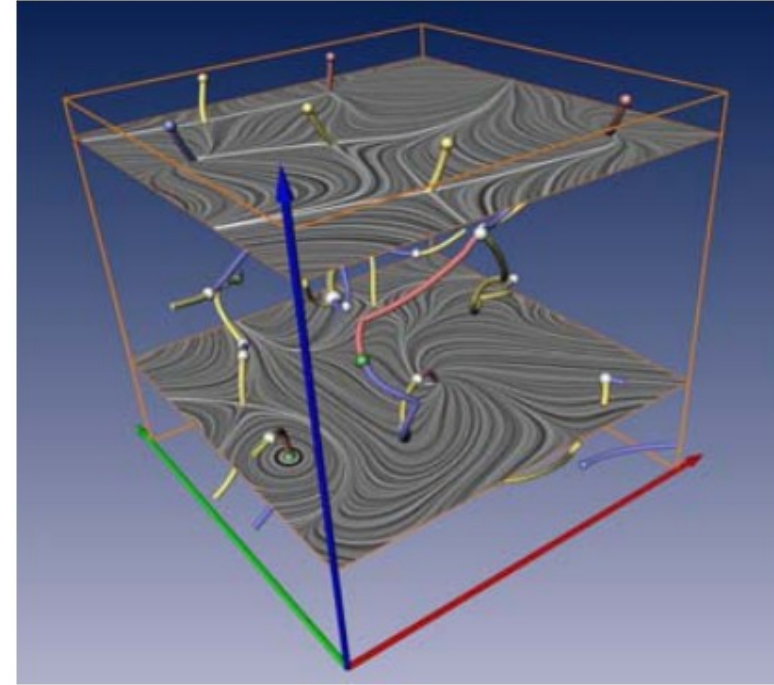
Introduction

Topological data analysis : Focuses on extracting critical points and how they relate to each other

Critical points : (Usually) locations of a function where gradient is 0

Singularities : Areas where the flow velocity is zero – e.g. sources, sinks

- Connectivity (*topology*) between the singularities is analyzed
- Topology of a vector field can be called "skeleton" of the flow
- Good for feature tracking

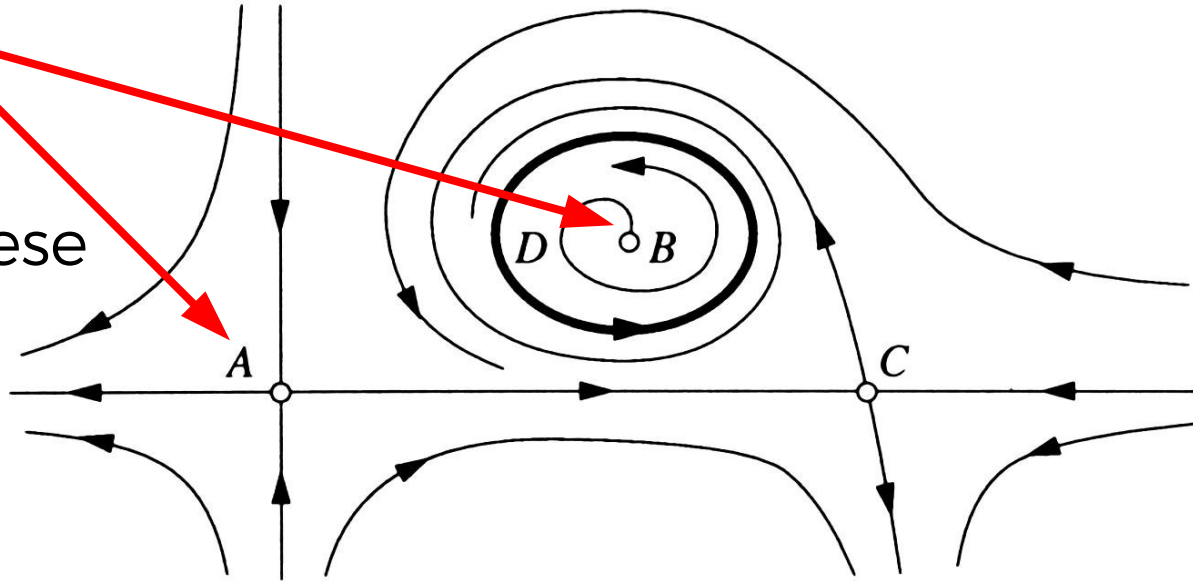


Theisel, H., T. Weinkauf, H.-C. Hege, and H.-P. Seidel.
"Stream Line and Path Line Oriented Topology for 2D
Time-Dependent Vector Fields." In IEEE Visualization
2004, 321-28, 2004.
<https://doi.org/10.1109/VISUAL.2004.99>.

Flow topology

- Critical points ($v(x) = 0$)
- Cycles ($s_x(t+T) = s_x(t)$)
- Connections between these

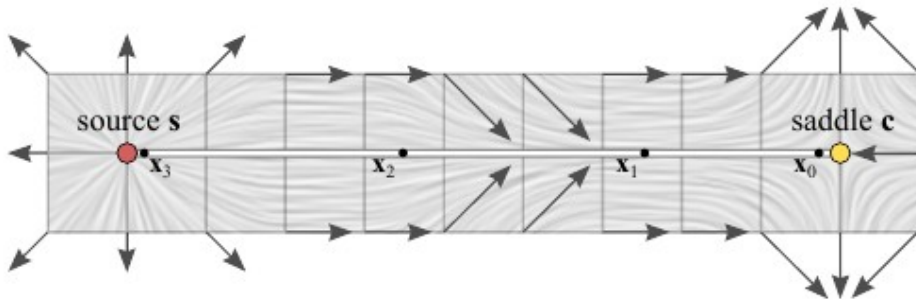
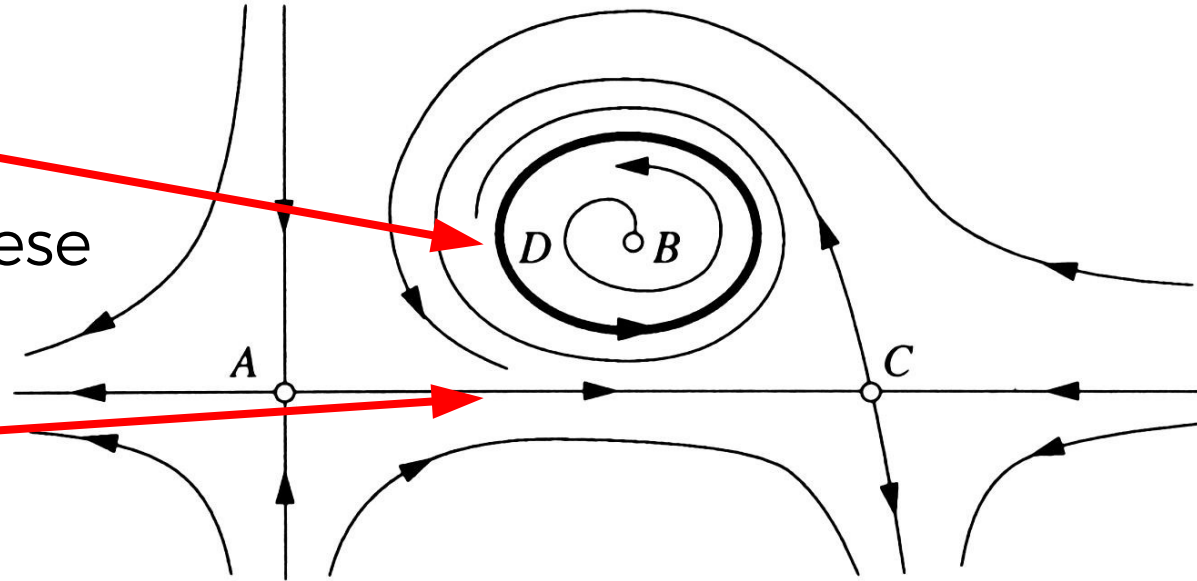
Separatrices : streamlines connecting critical points



Flow topology

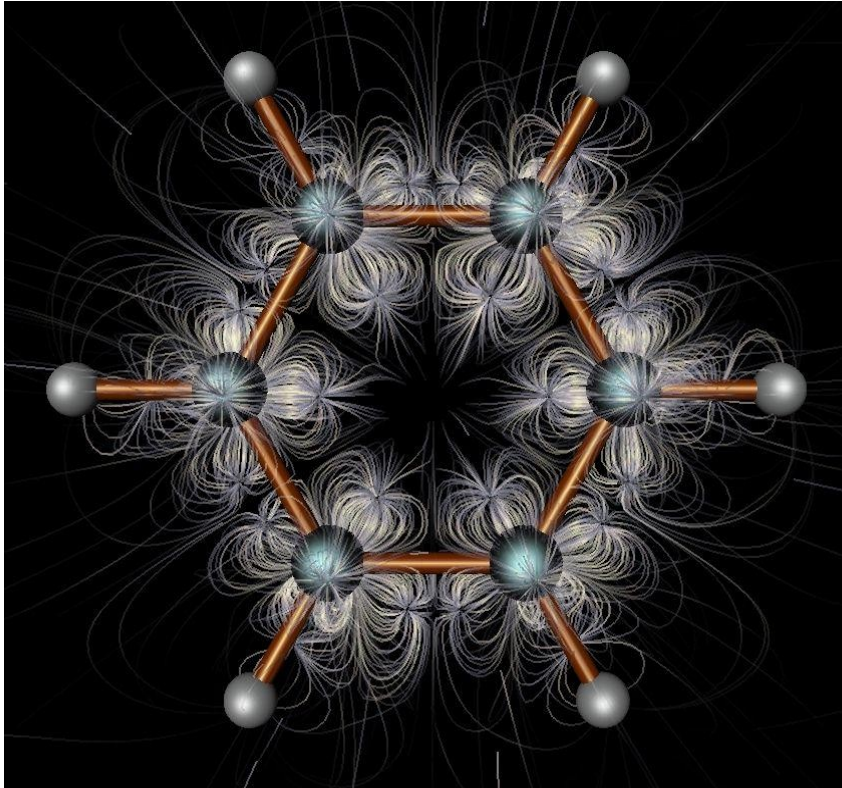
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Separatrices : streamlines connecting critical points

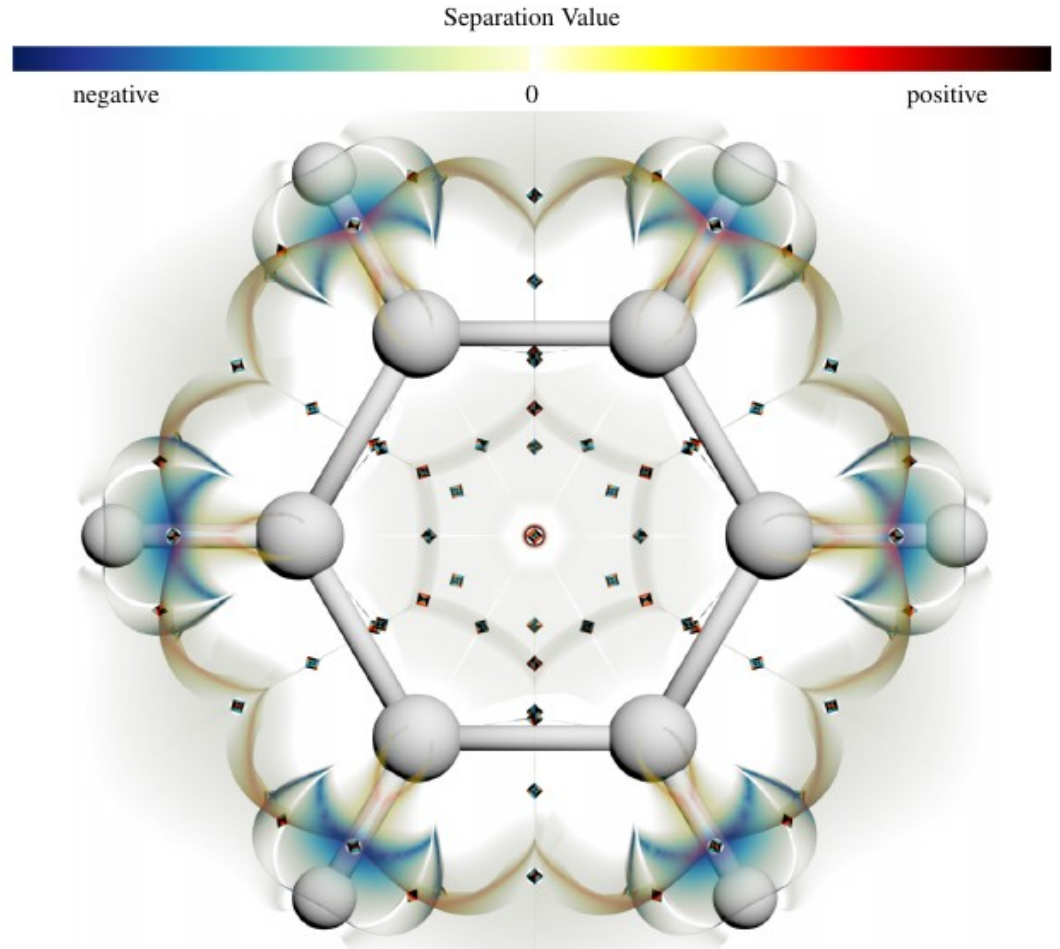


Hullin, Matthias, Reinhard Klein, Thomas Schultz, and Holger Theisel. "Finite Time Steady Vector Field Topology - Theoretical Foundation and 3D Case." The Eurographics Association, 2017.
<https://doi.org/10.2312/vmv.20171264>.

Example



VS



Hullin, Matthias, Reinhard Klein, Thomas Schultz, and Holger Theisel. "Finite Time Steady Vector Field Topology - Theoretical Foundation and 3D Case." The Eurographics Association, 2017. <https://doi.org/10.2312/vmv.20171264>.

Conclusion

Summary

- LIC – overview, easy to understand
- Feature-based – when you know what you're looking for
- Topology – hybrid method, good for feature tracking

Reading

- Hullin, Matthias, Reinhard Klein, Thomas Schultz, and Holger Theisel. “Finite Time Steady Vector Field Topology - Theoretical Foundation and 3D Case.” The Eurographics Association, 2017.
<https://doi.org/10.2312/vmv.20171264>.
- Han, Jun, Jun Tao, and Chaoli Wang. “FlowNet: A Deep Learning Framework for Clustering and Selection of Streamlines and Stream Surfaces.” IEEE Transactions on Visualization and Computer Graphics, 2018, 1-1. <https://doi.org/10.1109/TVCG.2018.2880207>.
- Hong, Fan, Chufan Lai, Hanqi Guo, Enya Shen, Xiaoru Yuan, and Sikun Li. “FLDA: Latent Dirichlet Allocation Based Unsteady Flow Analysis.” IEEE Transactions on Visualization and Computer Graphics 20, no. 12 (December 2014): 2545-54. <https://doi.org/10.1109/TVCG.2014.2346416>.
- Theisel, H., T. Weinkauff, H.-C. Hege, and H.-P. Seidel. “Stream Line and Path Line Oriented Topology for 2D Time-Dependent Vector Fields.” In IEEE Visualization 2004, 321-28, 2004.
<https://doi.org/10.1109/VISUAL.2004.99>.
- Laramée, Robert S., Helwig Hauser, Lingxiao Zhao, and Frits H. Post. “Topology-Based Flow Visualization, The State of the Art.” In Topology-Based Methods in Visualization, edited by Helwig Hauser, Hans Hagen, and Holger Theisel, 1-19. Mathematics and Visualization. Berlin, Heidelberg: Springer, 2007.
https://doi.org/10.1007/978-3-540-70823-0_1.