

Feature Flow Fields Tutorial

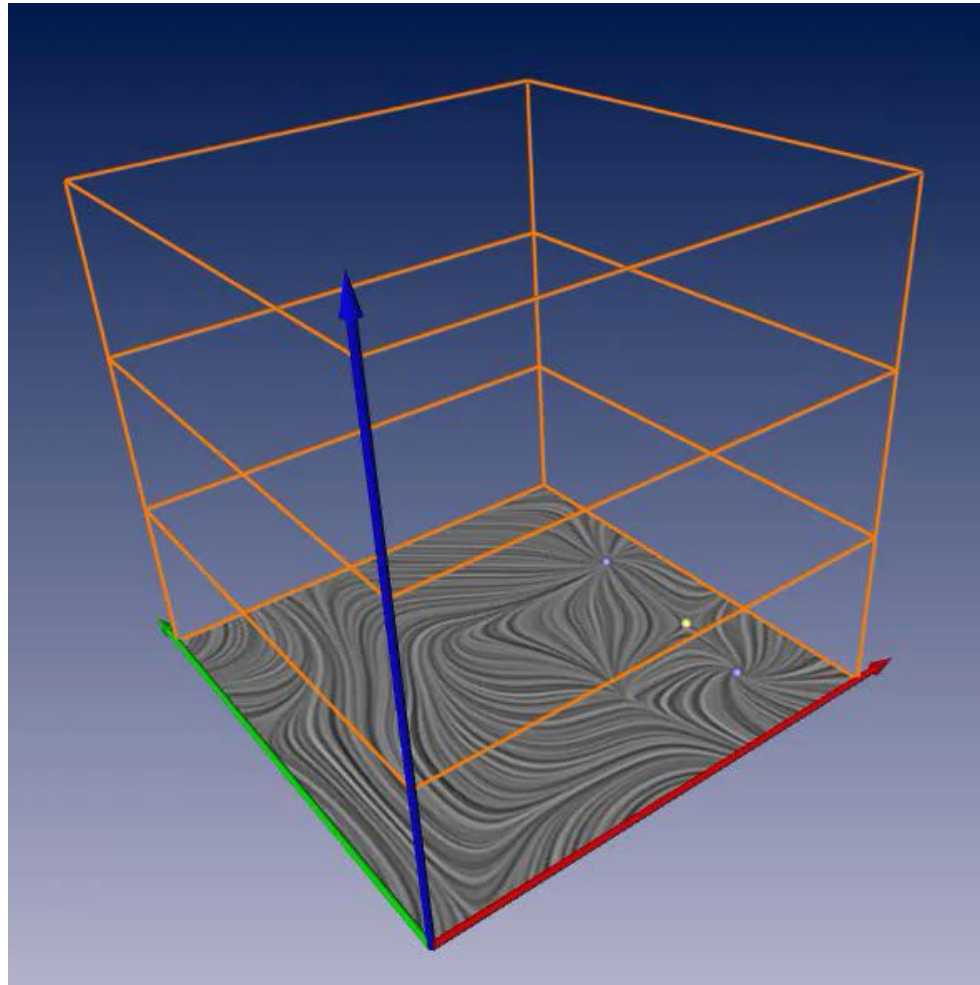
Speaker: Maik Schulze
University of Magdeburg, Germany



Visual Computing Research Group:
<http://vc.cs.ovgu.de/>

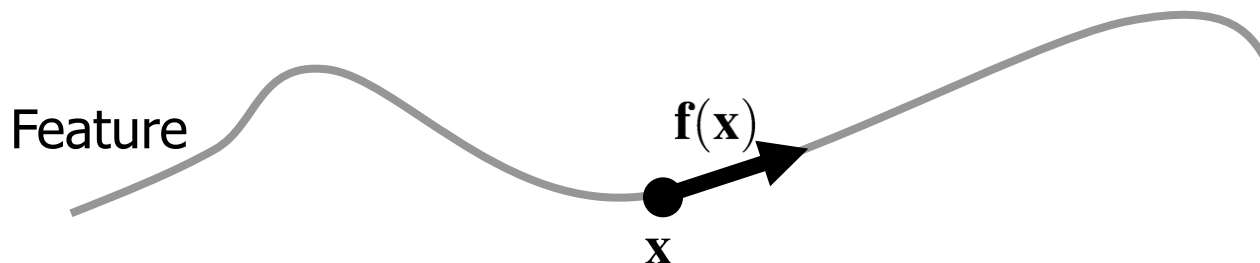
Based on slides from:
Holger Theisel
Tino Weinkauf

Motivation



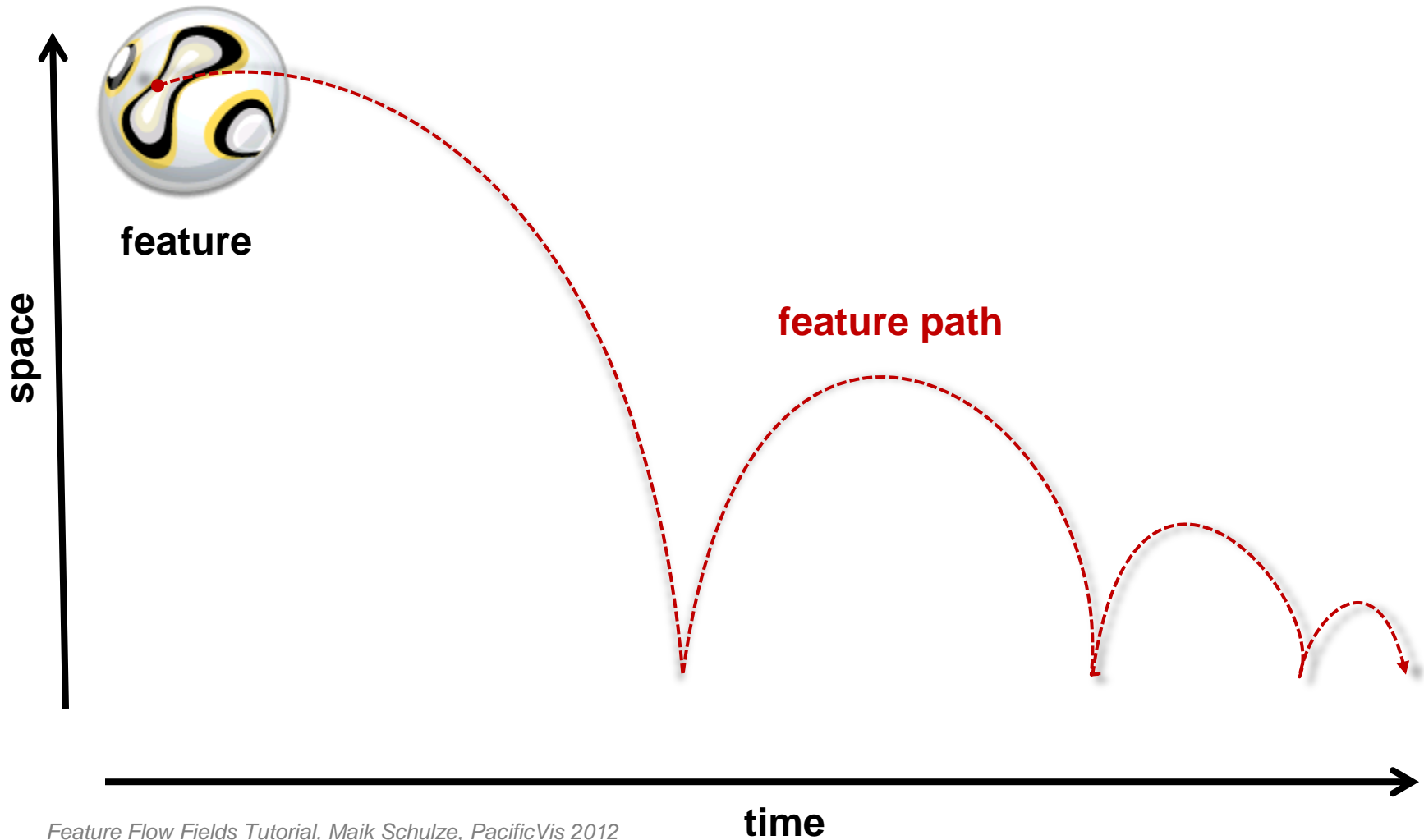
Motivation

- Capture topological information over time
 - Critical points
 - Periodic orbits
 - Vortex axes
- Data set can be scalar, vector or tensor field
- Represent dynamic behavior of features as integration in steady vector field
 - Stream lines, stream surfaces
 - Numerical stream line/surface integration is well-understood



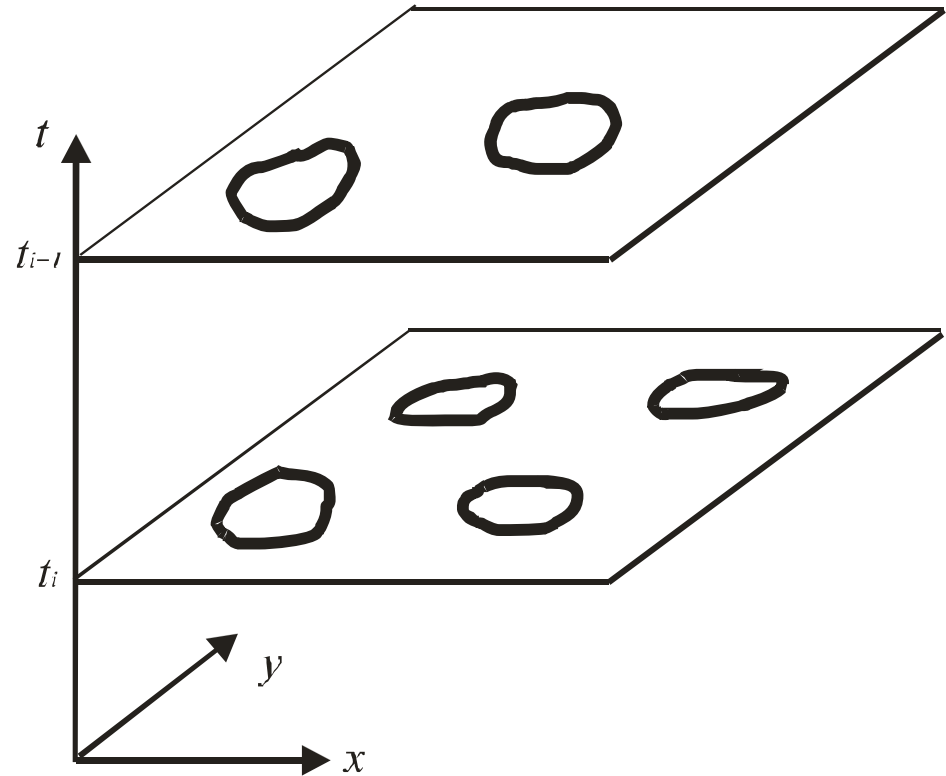
Feature Tracking

Simple Example: Tracking the ball



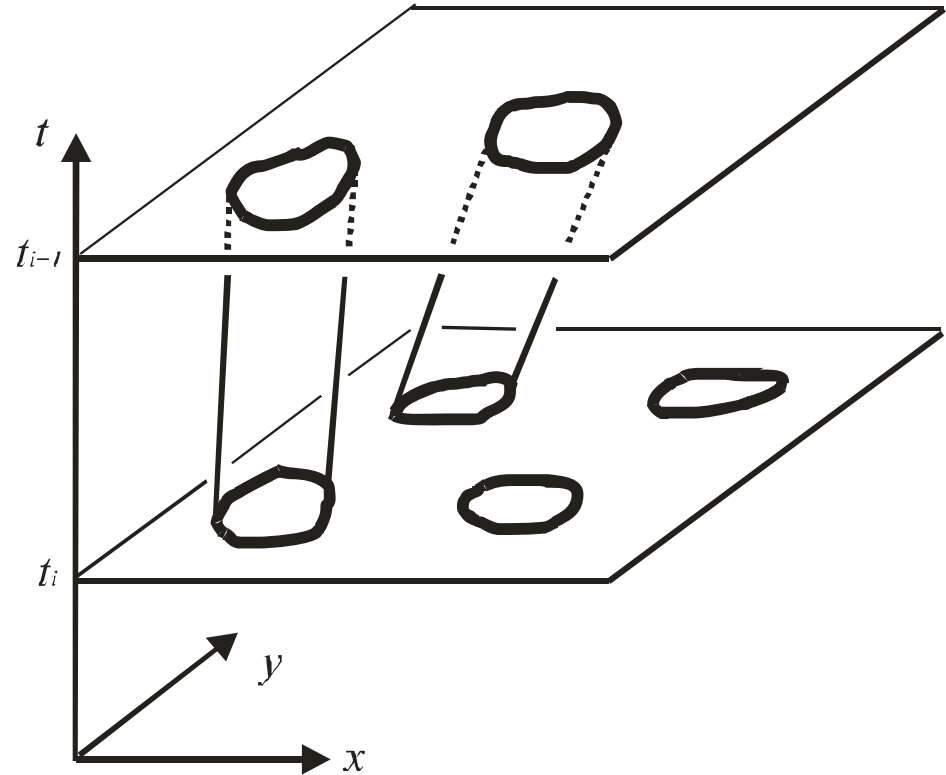
Feature Tracking

1. Extract features at different time steps



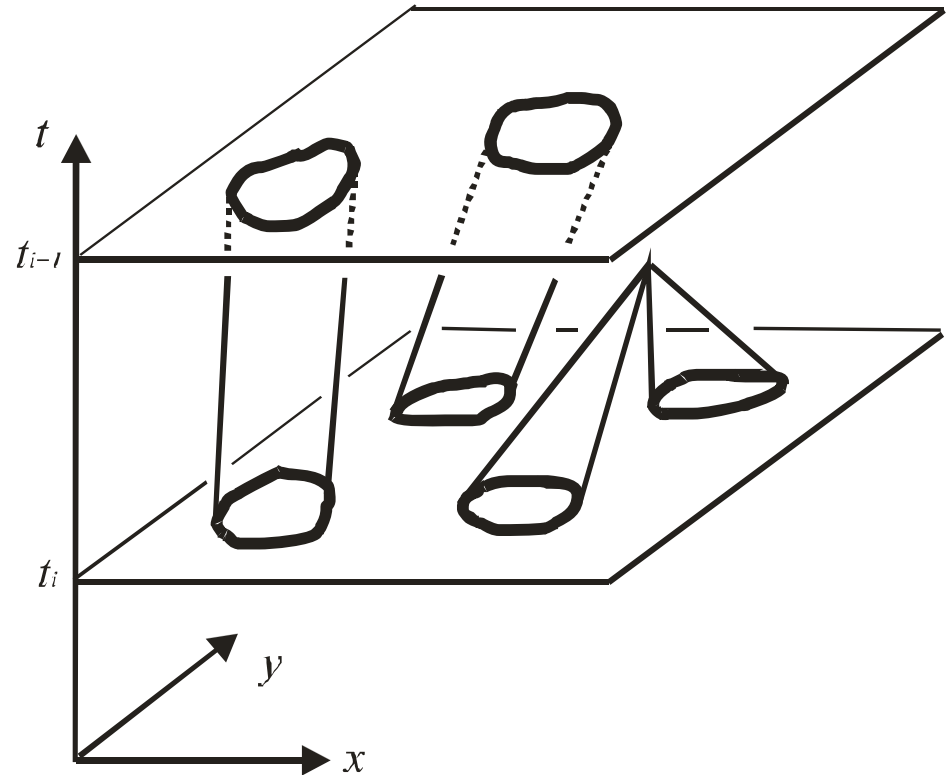
Feature Tracking

1. Extract features at different time steps
2. Find corresponding features



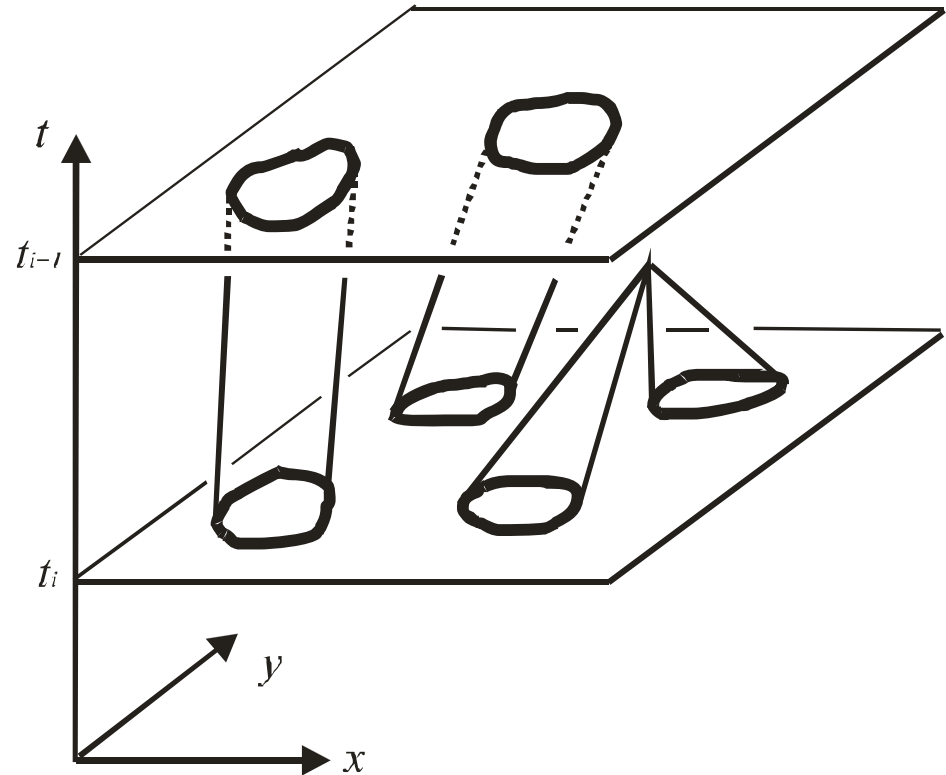
Feature Tracking

1. Extract features at different time steps
2. Find corresponding features
3. Event detection
 - birth, death
 - entry, exit
 - split, merge



Feature Tracking

1. Extract features at different time steps
2. Find corresponding features
3. Event detection
 - birth, death
 - entry, exit
 - split, merge
4. Visualization



Feature Tracking with FFF

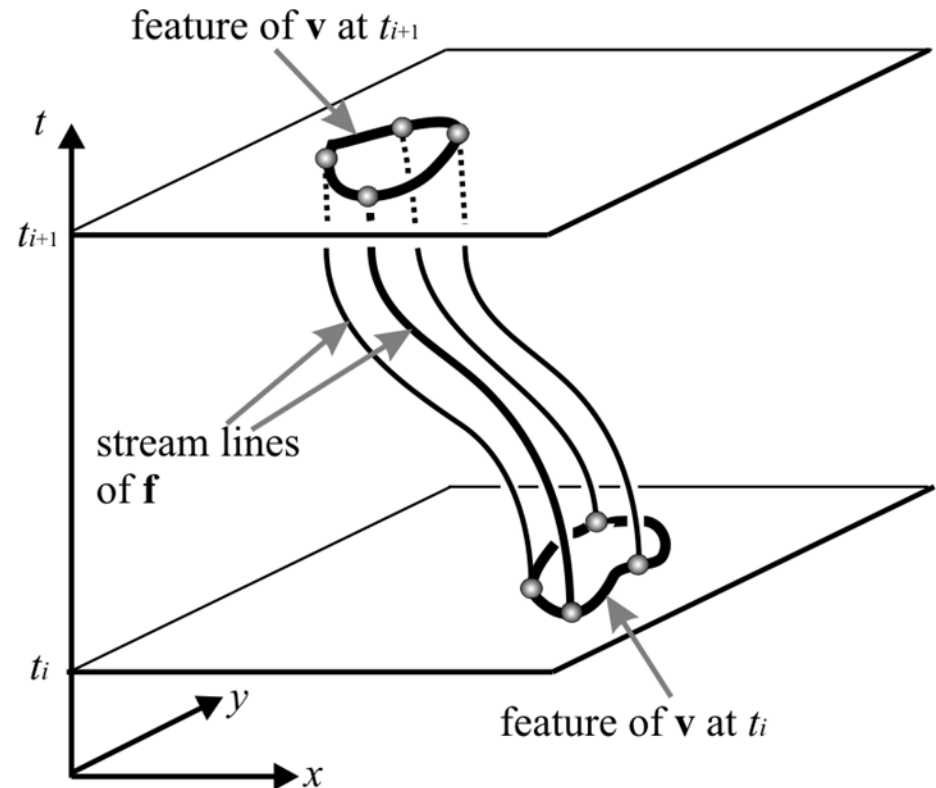
- Consider a point \mathbf{x} known to be part of a feature
 - FFF \mathbf{f} is a well-defined vector field at \mathbf{x}
 - \mathbf{f} points into direction where the feature moves to
 - Streamline integration in \mathbf{f} at \mathbf{x} yields curve with all points on the feature

$$\mathbf{v}(x, y, t) = \begin{pmatrix} u(x, y, t) \\ v(x, y, t) \end{pmatrix}$$



$$\mathbf{f}(x, y, t) = \begin{pmatrix} f(x, y, t) \\ g(x, y, t) \\ h(x, y, t) \end{pmatrix}$$

Feature Flow Field (FFF)



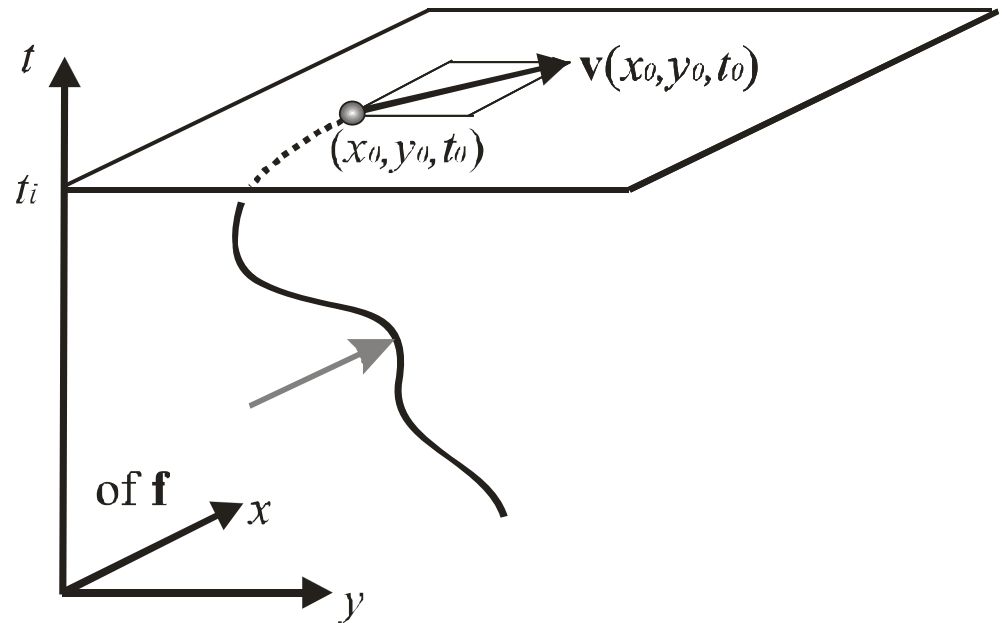
Application: Critical Point Tracking

Main Idea

$$\mathbf{v}(x, y, t) = \begin{pmatrix} u(x, y, t) \\ v(x, y, t) \end{pmatrix}$$



$$\mathbf{f}(x, y, t) = \begin{pmatrix} f(x, y, t) \\ g(x, y, t) \\ h(x, y, t) \end{pmatrix}$$



Feature flow field \mathbf{f} shall point into the direction where the original vector field \mathbf{v} is not changing in terms of both magnitude and direction.

Tracking Critical Points in 2D

$$\mathbf{v}(x, y, t) = \begin{pmatrix} u(x, y, t) \\ v(x, y, t) \end{pmatrix}$$

unsteady 2D vector field

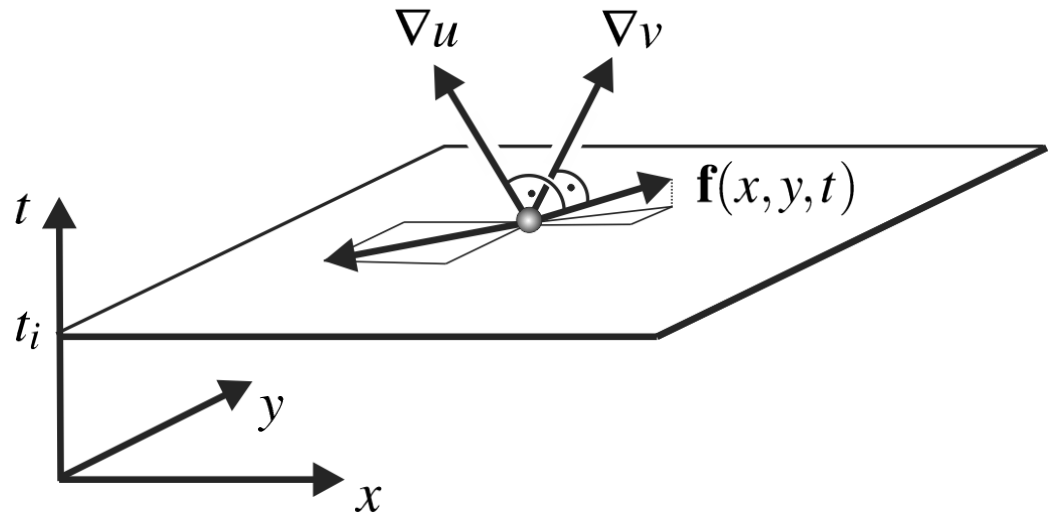


feature flow field

$$\mathbf{f}(x, y, t) = \begin{pmatrix} f(x, y, t) \\ g(x, y, t) \\ h(x, y, t) \end{pmatrix}$$

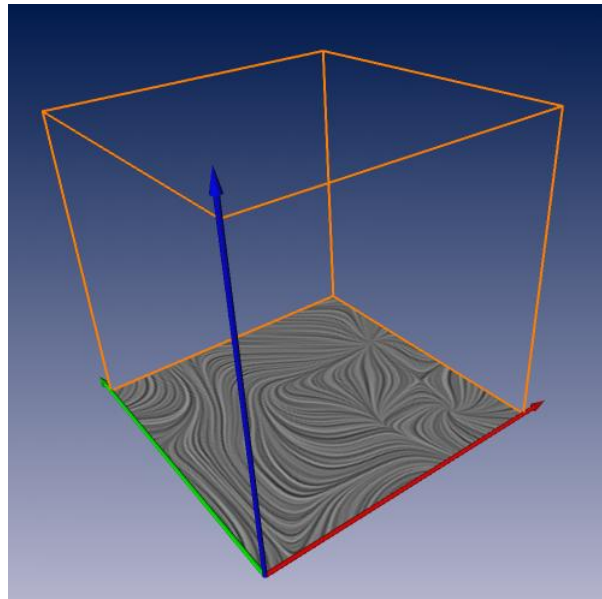


$$\mathbf{f}(x, y, t) = (\nabla u)^T \times (\nabla v)^T = \begin{pmatrix} \det(\mathbf{v}_y, \mathbf{v}_t) \\ \det(\mathbf{v}_t, \mathbf{v}_x) \\ \det(\mathbf{v}_x, \mathbf{v}_y) \end{pmatrix}$$

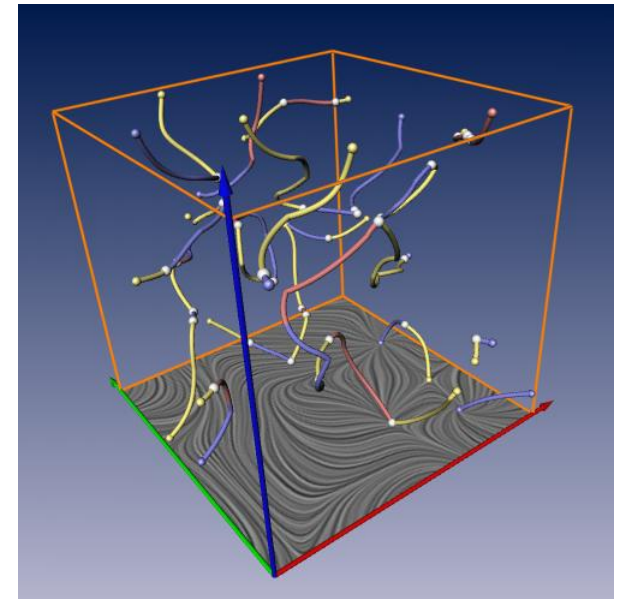


Tracking Critical Points in 2D

- Given: unsteady 2D field $\mathbf{v}(x, y, t) = \begin{pmatrix} u(x, y, t) \\ v(x, y, t) \end{pmatrix}$
- Goal: Find trajectories of all critical points



**FFF-based
Tracking**



Tracking Critical Points in 2D

1. Find all seed points

- Domain boundaries

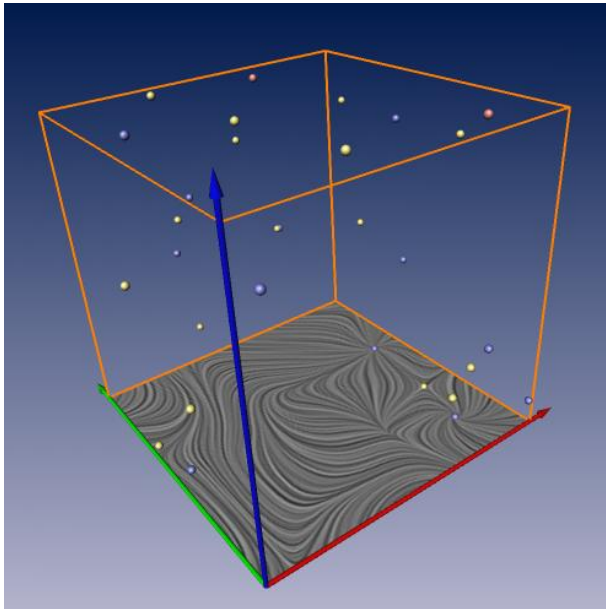
$$\mathbf{v}(x, y, t_{min}) = (0, 0)^T \text{ and } \mathbf{v}(x, y, t_{max}) = (0, 0)^T$$

$$\mathbf{v}(x, y_{min}, t) = (0, 0)^T \text{ and } \mathbf{v}(x, y_{max}, t) = (0, 0)^T$$

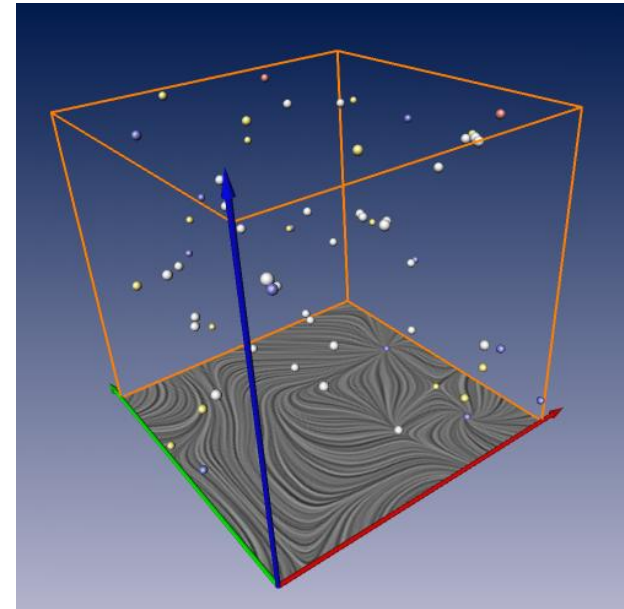
$$\mathbf{v}(x_{min}, y, t) = (0, 0)^T \text{ and } \mathbf{v}(x_{max}, y, t) = (0, 0)^T$$

- Fold bifurcations

$$[\mathbf{v}(\mathbf{x}) = (0, 0)^T, \det(\mathbf{J}_{\mathbf{v}}(\mathbf{x})) = 0]$$



On domain boundaries



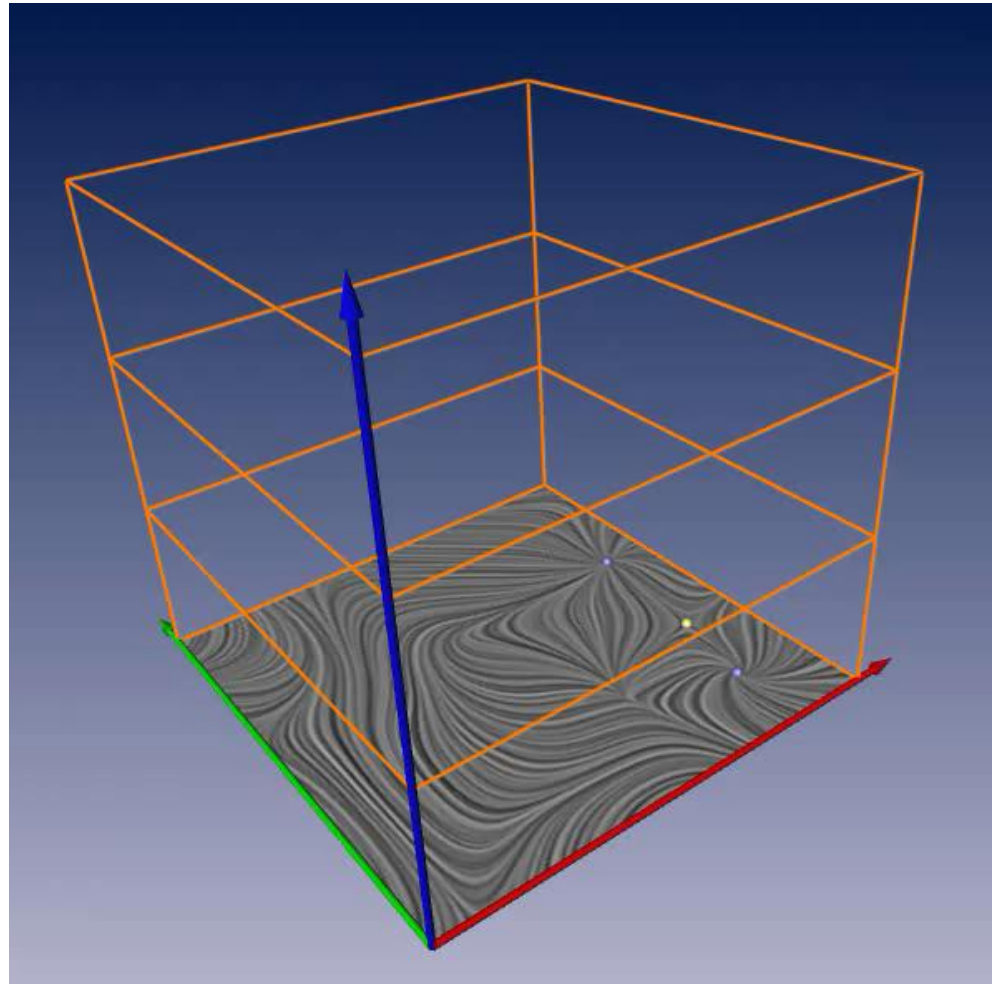
All seed points

Tracking Critical Points in 2D

$$\mathbf{v}(x, y, t) = \begin{pmatrix} u(x, y, t) \\ v(x, y, t) \end{pmatrix}$$

$$\mathbf{f}(x, y, t) = \begin{pmatrix} \det(\mathbf{v}_y, \mathbf{v}_t) \\ \det(\mathbf{v}_t, \mathbf{v}_x) \\ \det(\mathbf{v}_x, \mathbf{v}_y) \end{pmatrix}$$

- Optimization
 - One sweep through the data
 - Only two time slices at once in memory

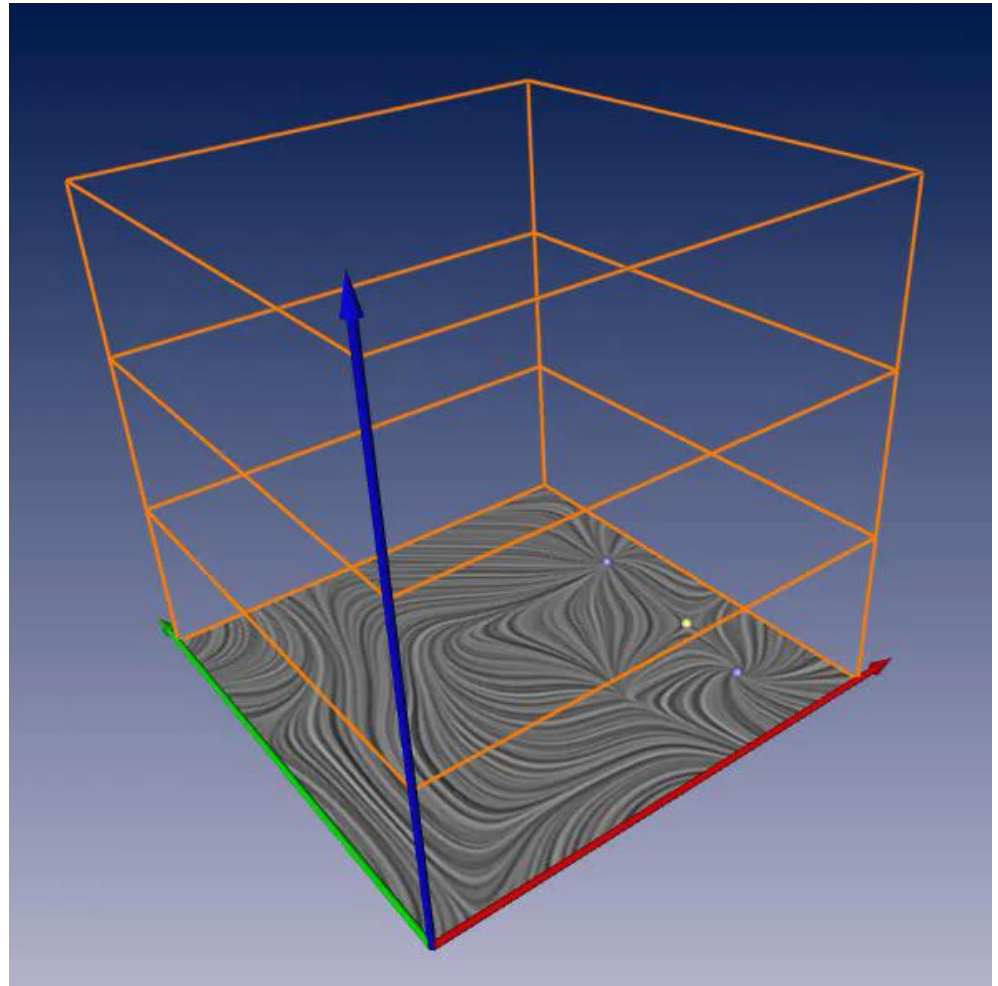


Tracking Critical Points in 2D

$$\mathbf{v}(x, y, t) = \begin{pmatrix} u(x, y, t) \\ v(x, y, t) \end{pmatrix}$$

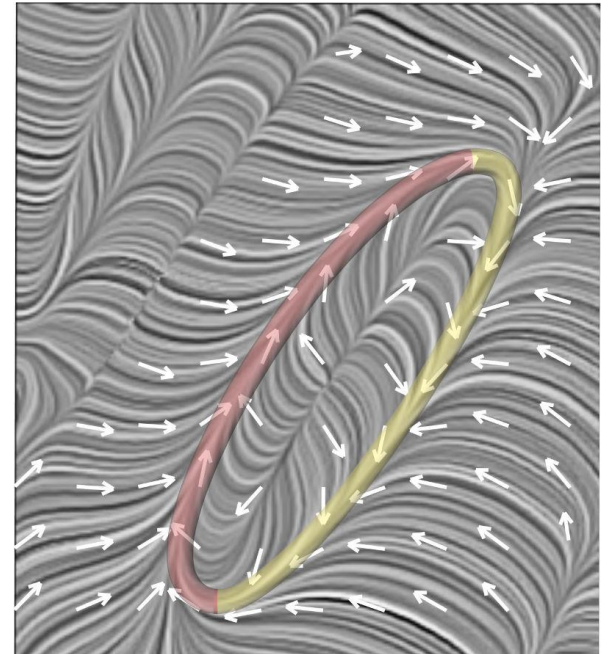
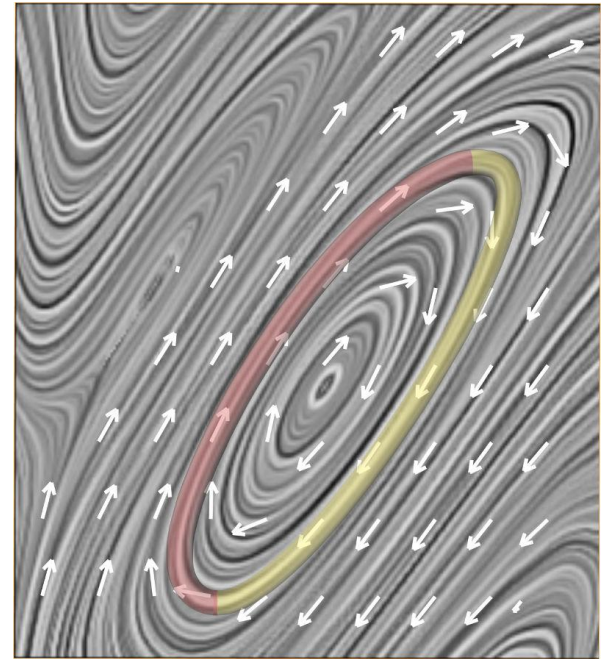
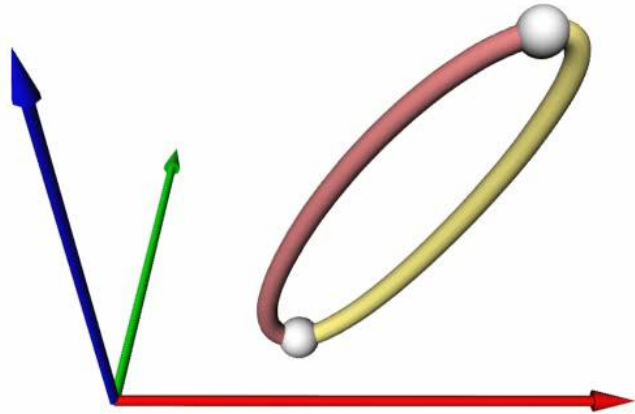
$$\mathbf{f}(x, y, t) = \begin{pmatrix} \det(\mathbf{v}_y, \mathbf{v}_t) \\ \det(\mathbf{v}_t, \mathbf{v}_x) \\ \det(\mathbf{v}_x, \mathbf{v}_y) \end{pmatrix}$$

- Optimization
 - One sweep through the data
 - Only two time slices at once in memory



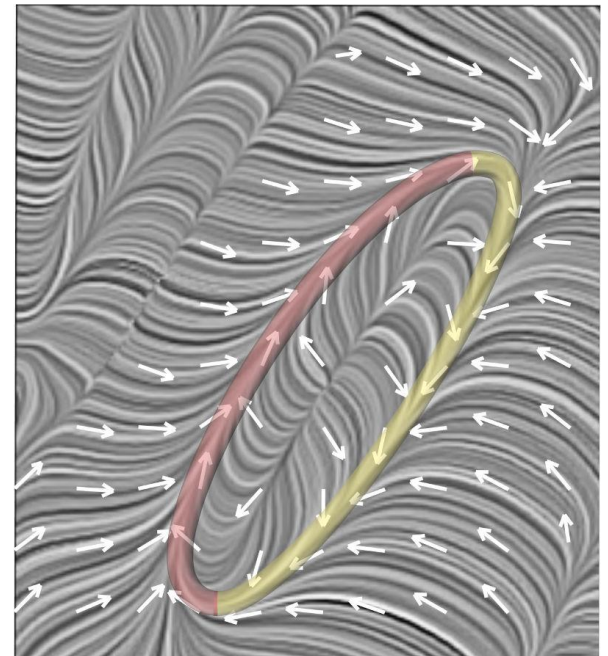
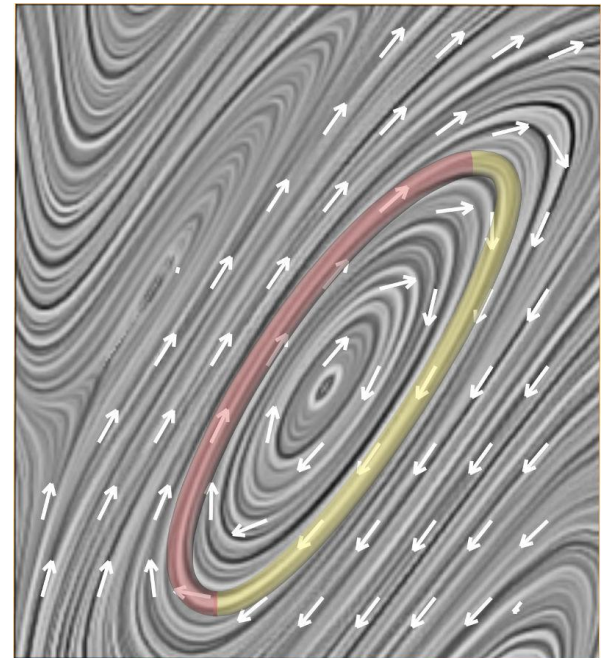
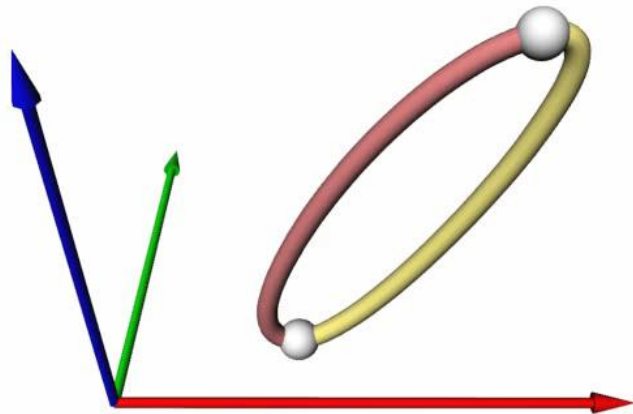
Stability of FFF

- Streamlines in f have possibly diverging behavior
- In long integration times, streamline diverges from actual feature line



Stability of FFF

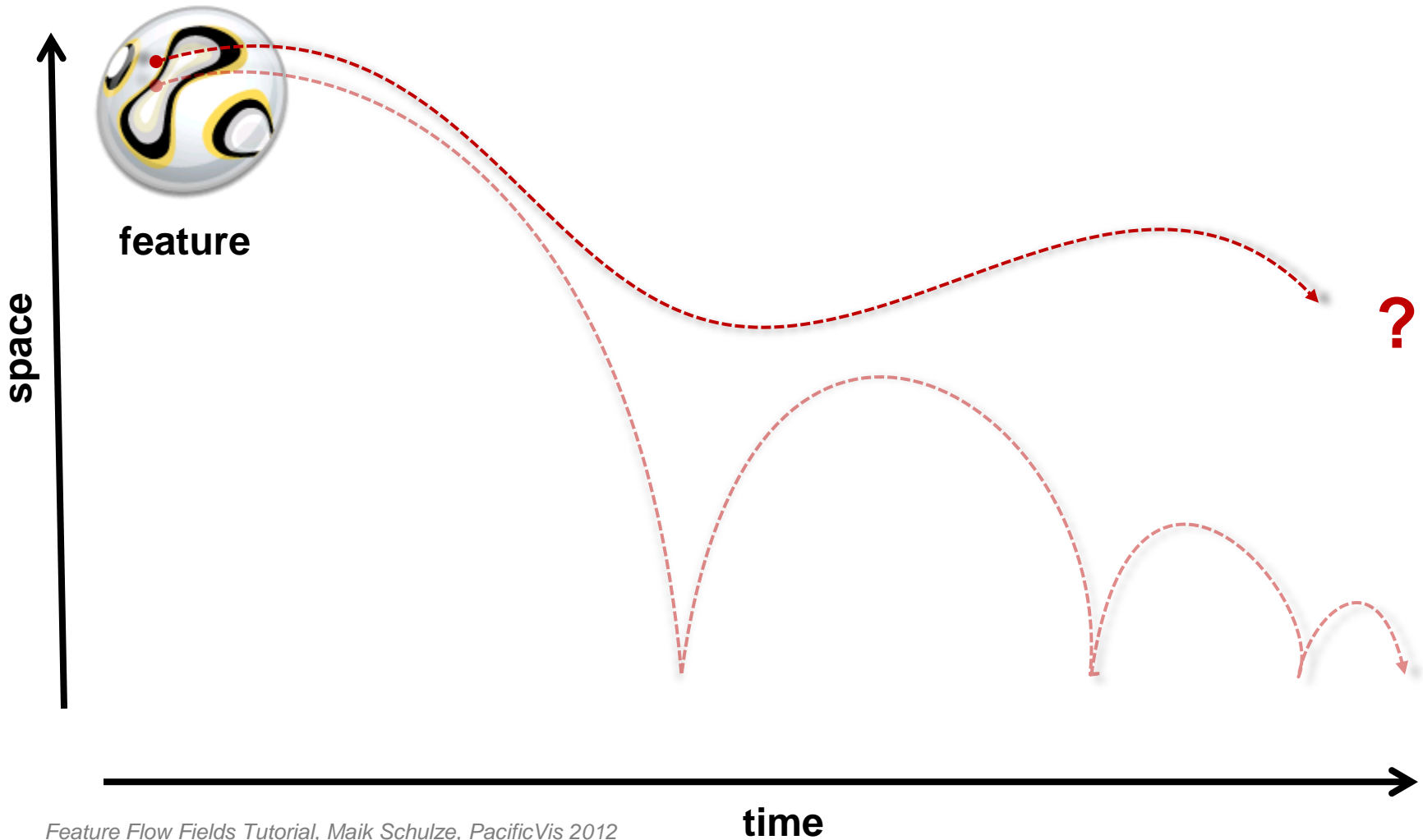
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Stable FFF

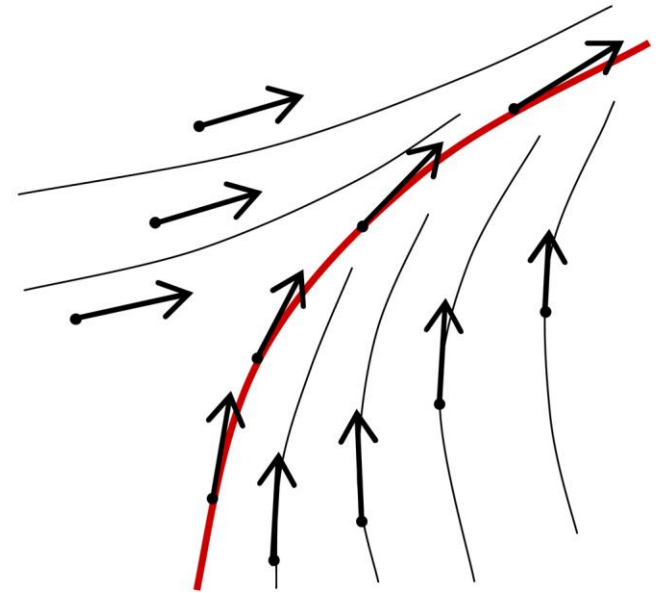
Feature Tracking

Simple Example: Tracking the ball

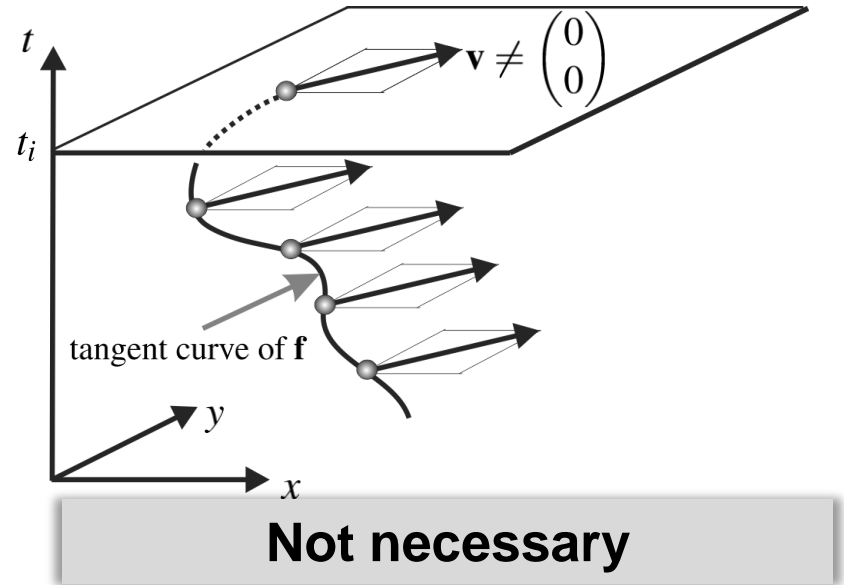
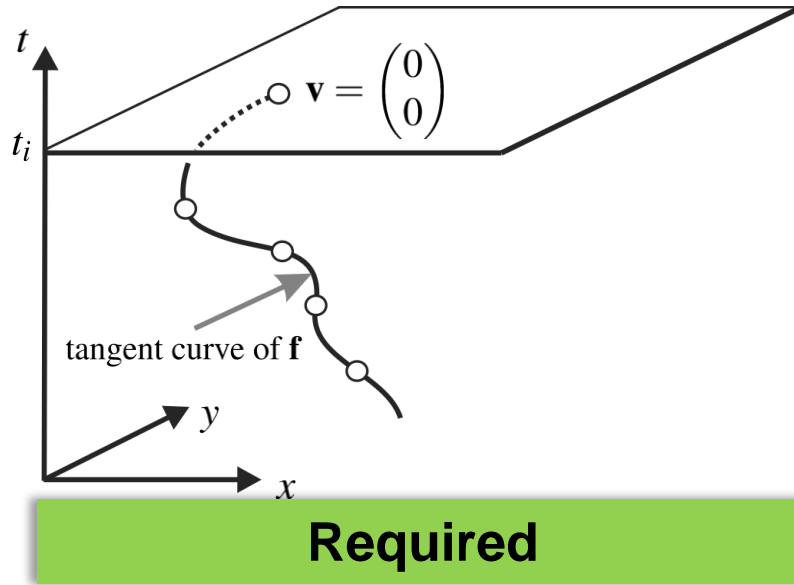


Stable Feature Flow Fields

- Mathematically correct
- Numerically stable
 - Converging flow behavior near the feature line
 - Automatic correction of small errors
- Feature tracking still stream line integration
- Seeding must not be as precisely computed as in FFF

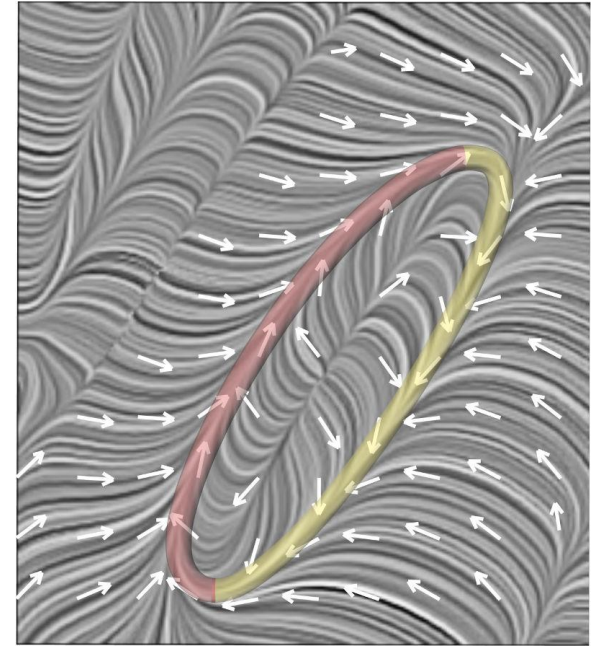
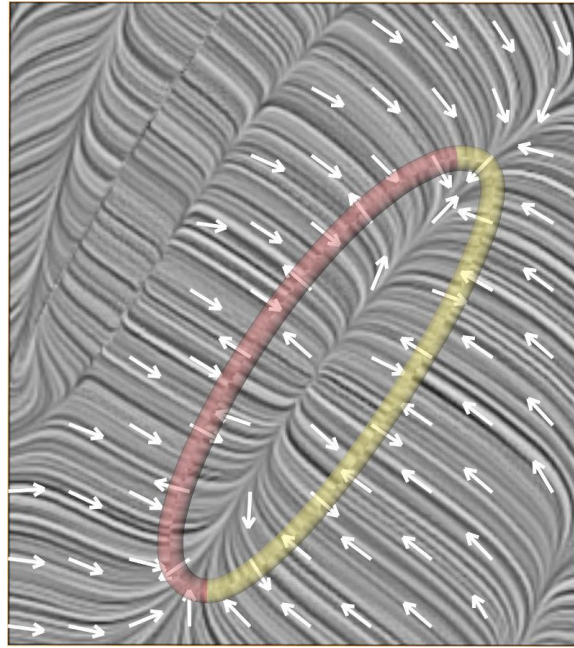
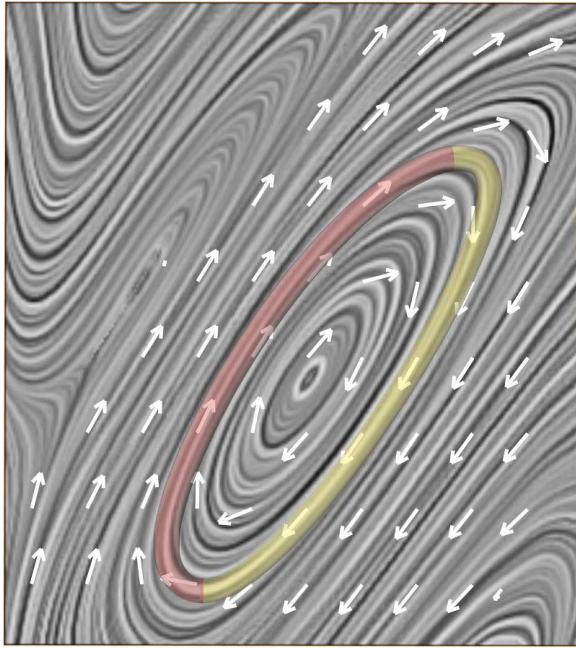


Stable Tracking Critical Points in 2D



$$\mathbf{f}(x, y, t) = (\nabla u)^T \times (\nabla v)^T = \begin{pmatrix} \det(\mathbf{v}_y, \mathbf{v}_t) \\ \det(\mathbf{v}_t, \mathbf{v}_x) \\ \det(\mathbf{v}_x, \mathbf{v}_y) \end{pmatrix}$$

Stable Tracking Critical Points in 2D



$$\mathbf{f}(x, y, t) = (\nabla u)^T \times (\nabla v)^T$$

$$\mathbf{g} = \frac{\mathbf{f}}{\|\mathbf{f}\|} \times \begin{pmatrix} \det(\mathbf{v}, \mathbf{v}_x) \\ \det(\mathbf{v}, \mathbf{v}_y) \\ \det(\mathbf{v}, \mathbf{v}_t) \end{pmatrix}$$

$$\mathbf{h} = \mathbf{f} + \alpha \mathbf{g}$$

acts as sink around $\mathbf{v} = \mathbf{0}$

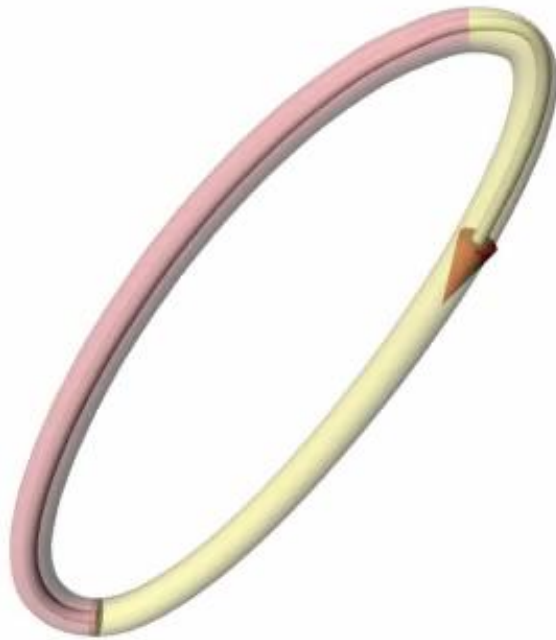
Original FFF

$$\mathbf{v} = \mathbf{0} \Rightarrow \mathbf{g} = \mathbf{0}$$

Stable Feature Flow Field

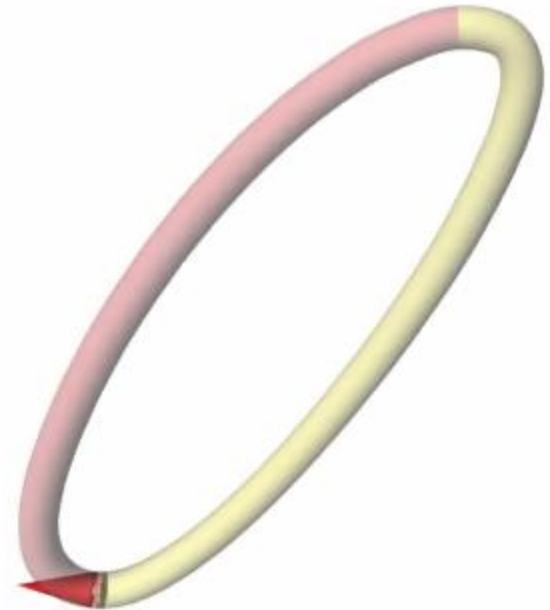
Stable Tracking Critical Points in 2D

216 integration steps



Feature Flow Fields

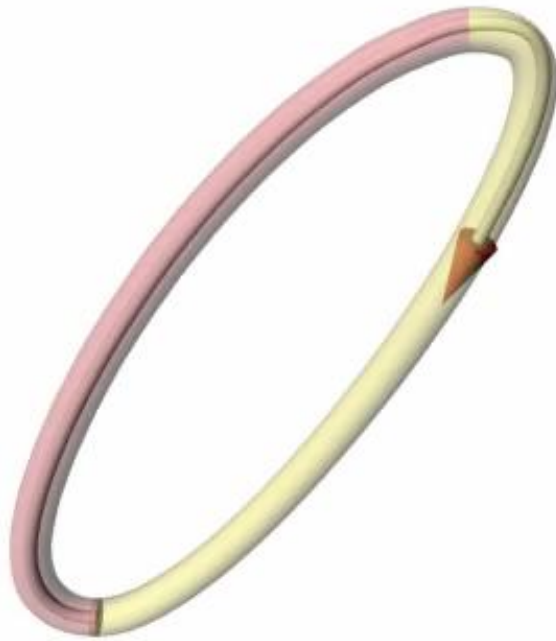
2 integration steps



Stable Feature Flow Fields

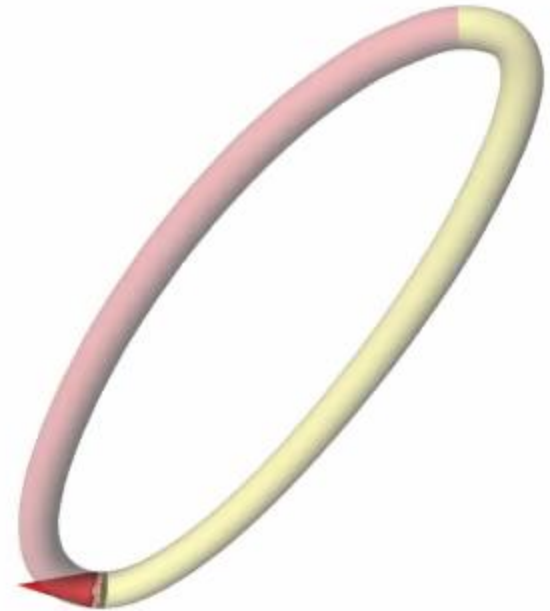
Stable Tracking Critical Points in 2D

216 integration steps



Feature Flow Fields

2 integration steps



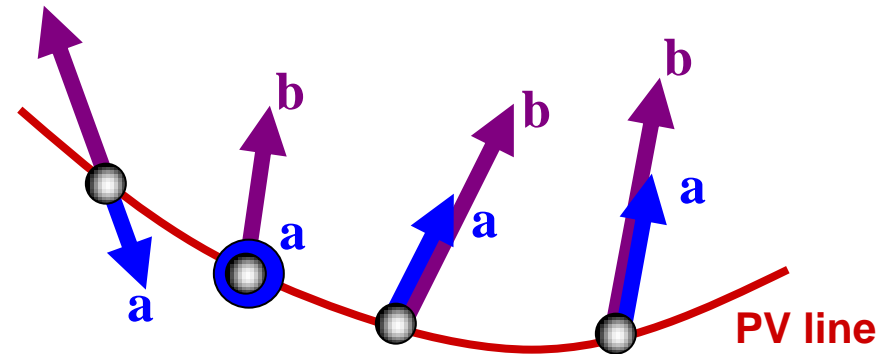
Stable Feature Flow Fields

Stable Feature Flow Fields for Parallel Vector Lines

Parallel Vectors with Stable FFF

- **Parallel Vectors (PV) operator**
 - Generic approach to feature extraction
 - Swirling motion cores, ridges, ...

$$\mathbf{w}_1 \parallel \mathbf{w}_2$$



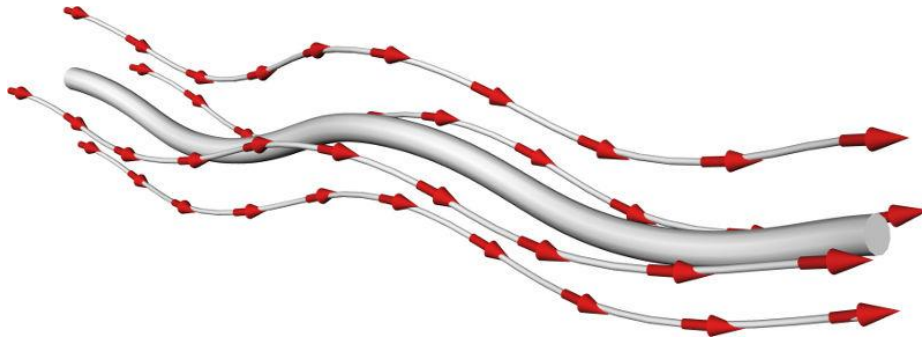
- Every PV problem can be reformulated using FFF

$$\mathbf{f} = \begin{pmatrix} \det(\mathbf{q}_y, \mathbf{q}_z, \mathbf{a}) \\ \det(\mathbf{q}_z, \mathbf{q}_x, \mathbf{a}) \\ \det(\mathbf{q}_x, \mathbf{q}_y, \mathbf{a}) \end{pmatrix} \quad \text{with} \quad \mathbf{q} = \mathbf{w}_1 \times \mathbf{w}_2$$

Parallel Vectors with Stable FFF

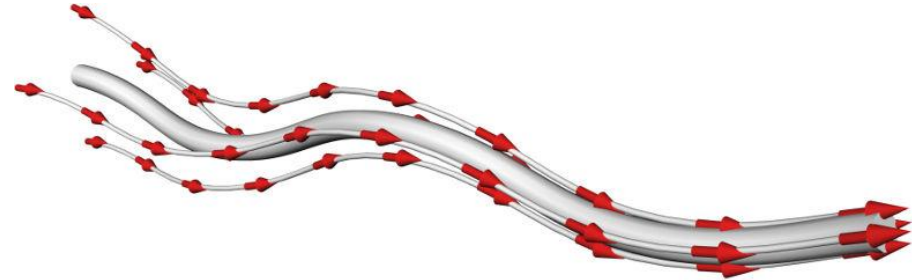
Correction Field

$$\mathbf{g} = \frac{\mathbf{f}}{\|\mathbf{f}\|} \times \begin{pmatrix} \det(\mathbf{q}, \mathbf{q}_x, \mathbf{a}) \\ \det(\mathbf{q}, \mathbf{q}_y, \mathbf{a}) \\ \det(\mathbf{q}, \mathbf{q}_z, \mathbf{a}) \end{pmatrix}$$



$$\mathbf{f} = \begin{pmatrix} \det(\mathbf{q}_y, \mathbf{q}_z, \mathbf{a}) \\ \det(\mathbf{q}_z, \mathbf{q}_x, \mathbf{a}) \\ \det(\mathbf{q}_x, \mathbf{q}_y, \mathbf{a}) \end{pmatrix}$$

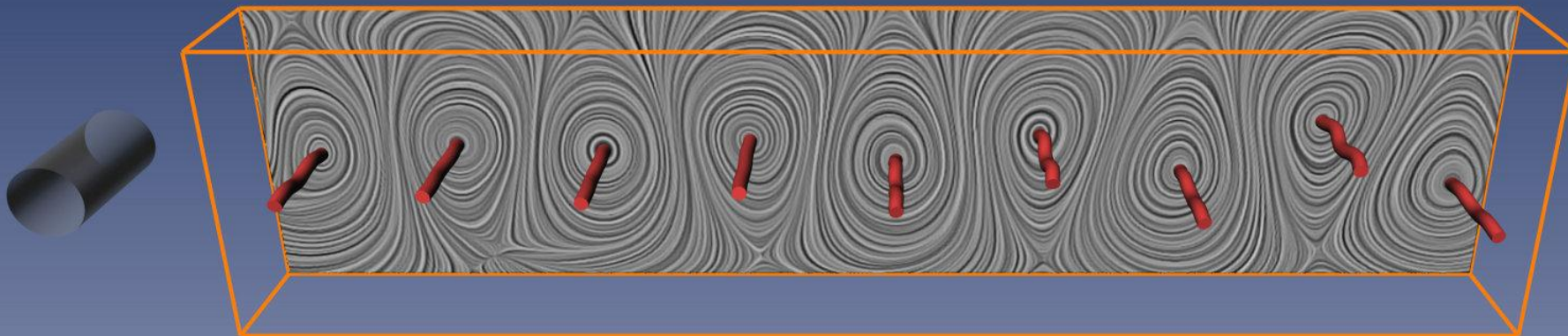
Original FFF



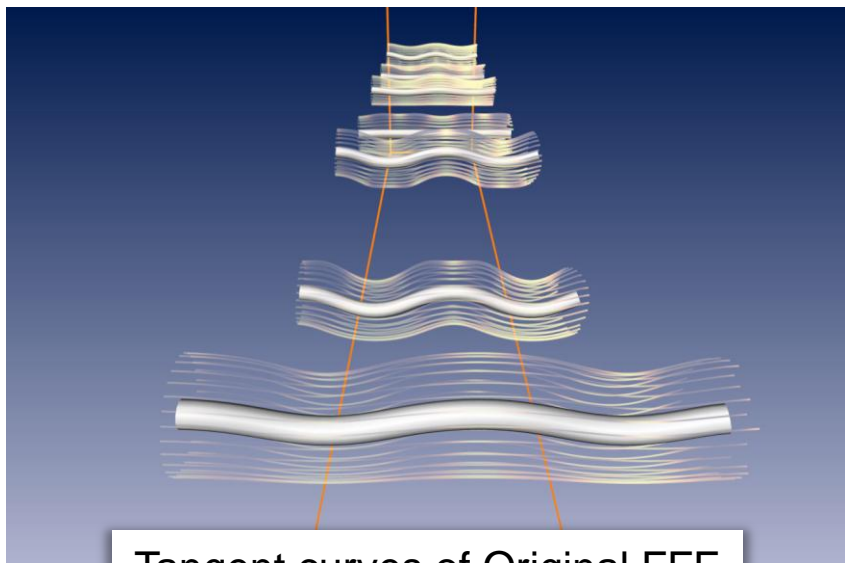
$$\mathbf{h} = \mathbf{f} + \alpha \mathbf{g}$$

Stable Feature Flow Field

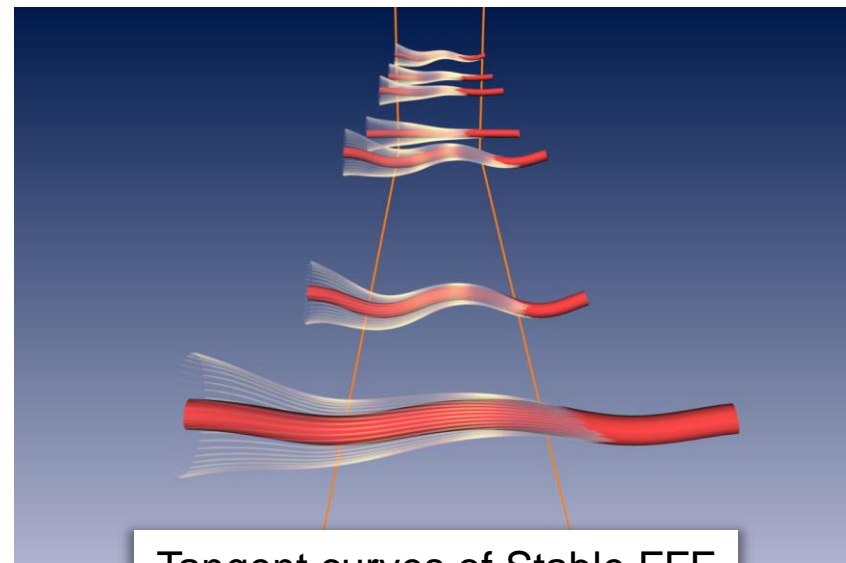
Parallel Vectors with Stable FFF



Vortex core lines (Sujudi/Haimes) extracted using Stable FFF



Tangent curves of Original FFF



Tangent curves of Stable FFF

Summary

- Dynamic behavior of certain features can be described by streamline integration in Feature Flow Fields
 - Critical points
 - Periodic orbits
 - Vortex axes

- Numerical stability can be improved by Stable Feature Flow Fields

Feature Flow Fields Tutorial

Speaker: Maik Schulze
University of Magdeburg, Germany

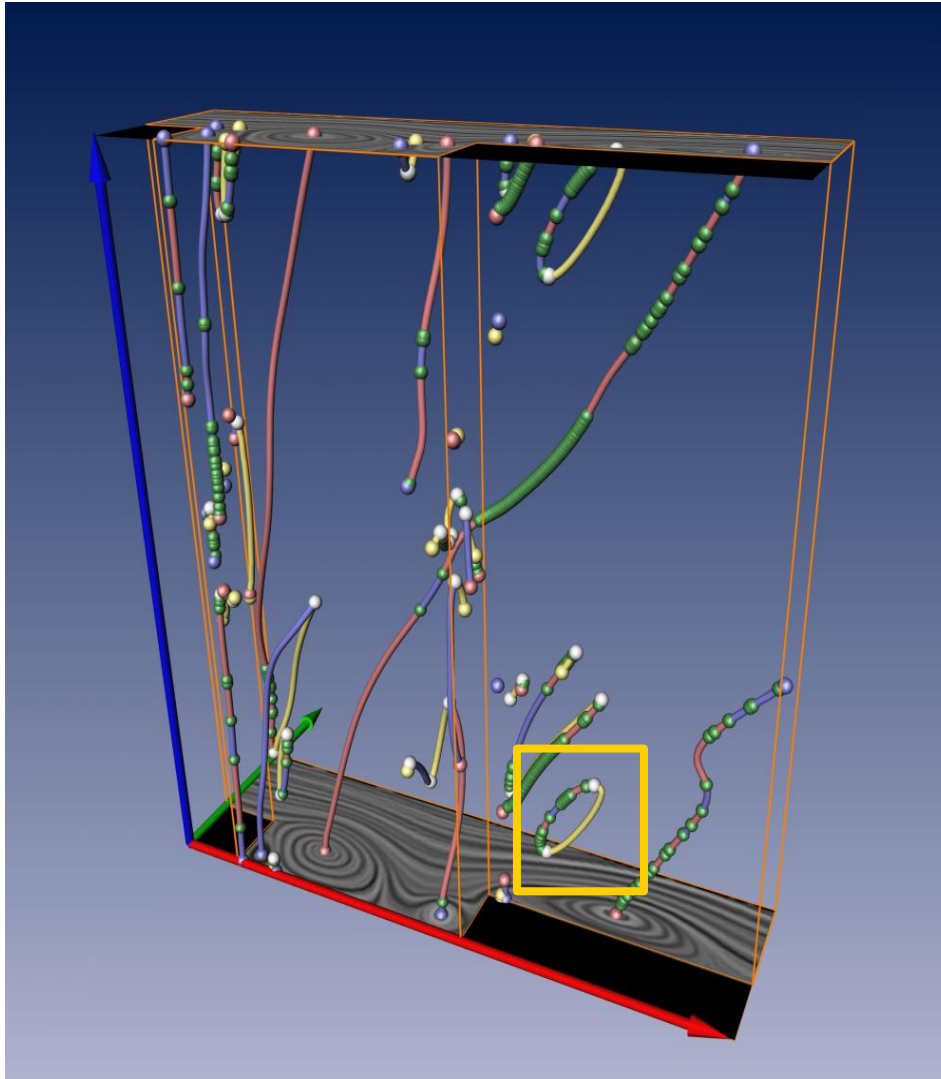


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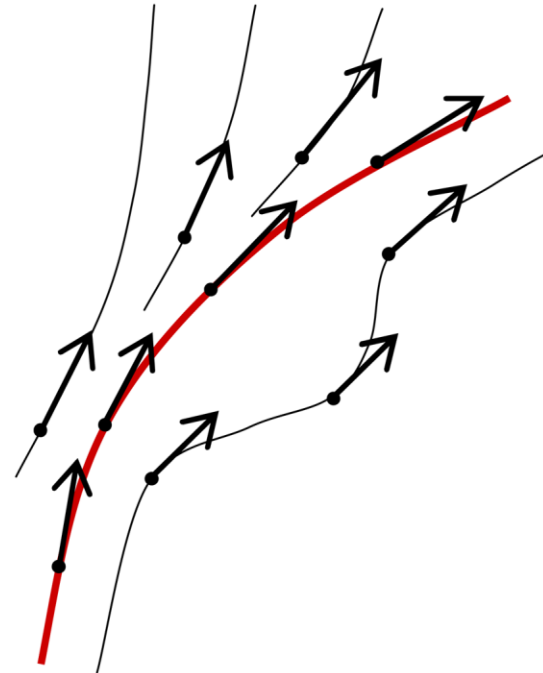
Stability of FFF

Stability of FFF



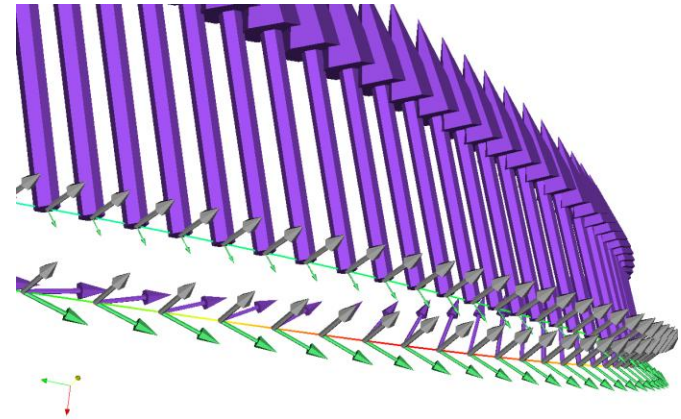
Stability of FFF

- Mathematically correct
- Possibly numerically unstable
 - Possibly divergent behavior near feature line
 - Numerical errors may move integration off the feature



Improvement of FFF

- Predictor-Corrector approach
 - Do integration step
 - Move particle to feature by root finding

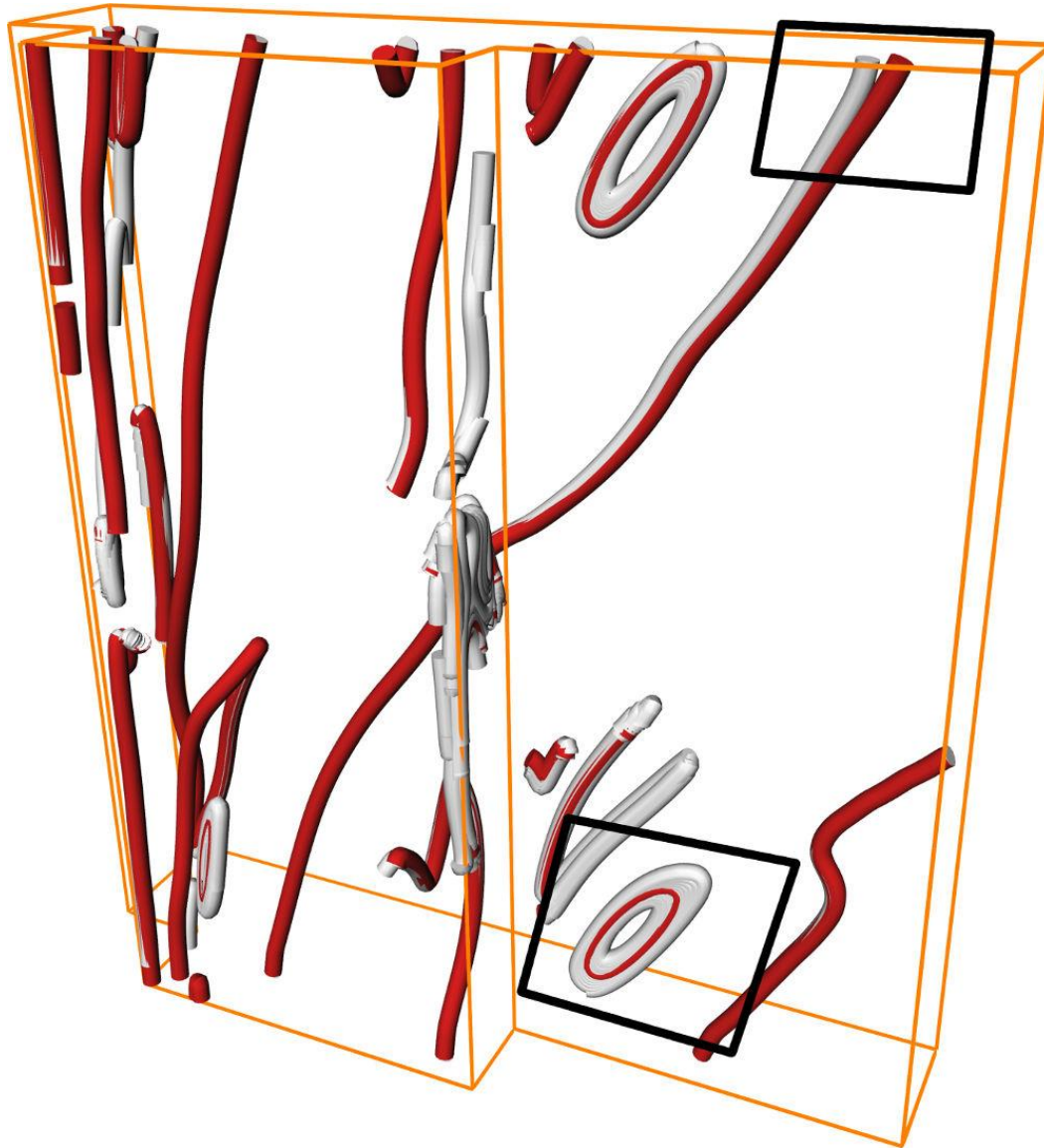


A. Van Gelder, A. Pang

Using PVsolve to Analyze and Locate Positions of Parallel Vectors,
IEEE TVCG 15(4):682-695, 2009.

- Disadvantage
 - Root finding computationally expensive
 - Feature tracking not stream line integration anymore

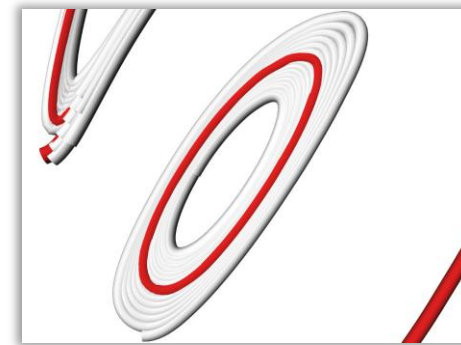
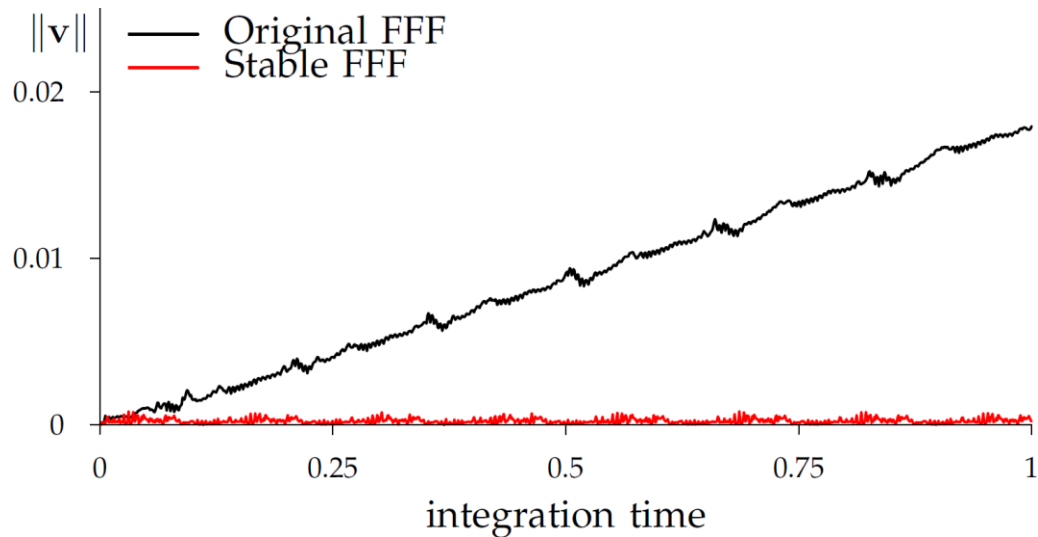
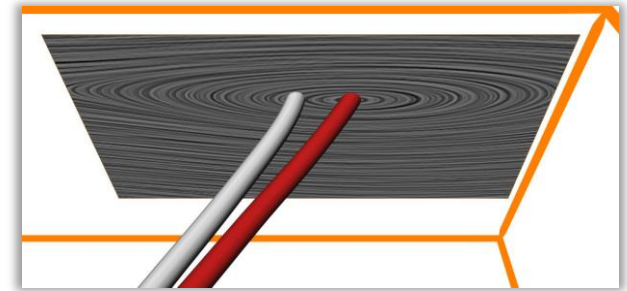
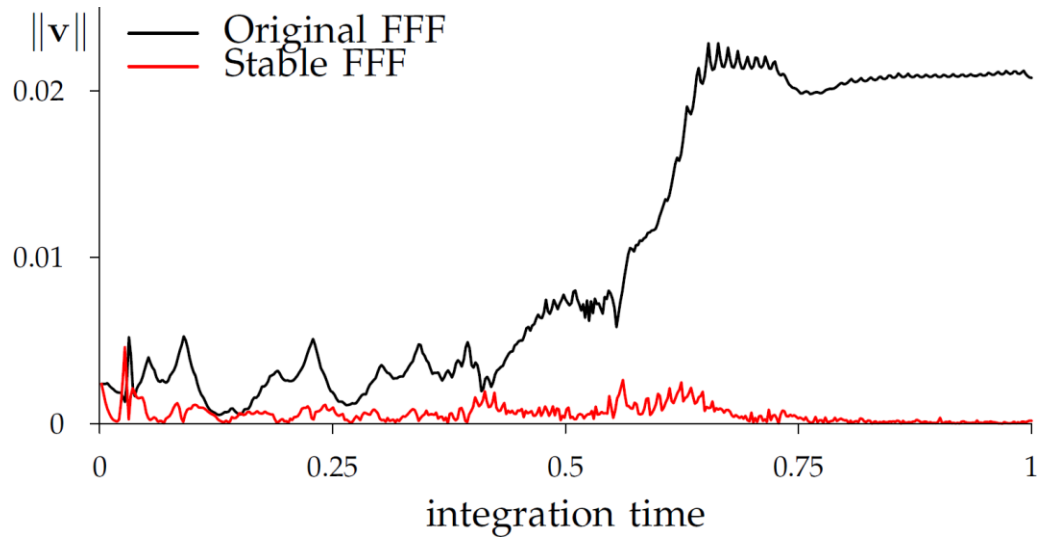
Comparison between FFF and Stable FFF



Original FFF

Stable FFF

Comparison between FFF and Stable FFF



Acknowledgements

- **Based on the following references:**

- **Scientific Visualization** **Tino Weinkauf**, MPI Saarbrücken, 2012
- **Flow and Tensor Visualization** **Holger Theisel**, University of Magdeburg, 2011
- **Flow Visualization** **Helwig Hauser**, University of Bergen, 2011

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Selected Work From Feature Flow Fields

- **H. Theisel and H.-P. Seidel**
Feature Flow Fields, VisSym 2003
- **H. Theisel, T. Weinkauff, H.-C. Hege, and H.-P. Seidel**
Stream Line and Path Line Oriented Topology for 2D Time-Dependent Vector Fields, IEEE Vis 2004
- **T. Weinkauff, H. Theisel, H.-C. Hege, and H.-P. Seidel**
Feature Flow Fields in Out-Of-Core Settings, TopoInVis 2005
- **X. Zheng and A. Pang**
Topological Lines in 3D Tensor Fields and Discriminant Hessian Factorization, TVCG 2005
- **H. Theisel, J. Sahner, T. Weinkauff, H.-C. Hege, and H.-P. Seidel**
Extraction of Parallel Vector Surfaces in 3D Time-Dependent Fields and Application to Vortex Core Line Tracking, IEEE Vis 2005
- **S. Depardon, J. J. Lasserre, L. E. Brizzi, and J. Borée**
Automated topology classification method for instantaneous velocity fields, Experiments in Fluids 2007
- **T. Weinkauff and H. Theisel and A. Van Gelder and A. Pang**
Stable Feature Flow Fields, TVCG 2010
- **J. Reininghaus, J. Kasten, T. Weinkauff, I. Hotz**
Efficient Computation of Combinatorial Feature Flow Fields, TVCG 2011