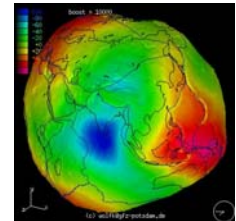
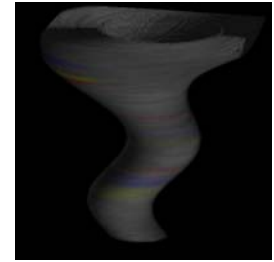


## Chapter 6 – Visualization Techniques for Vector Fields

- 6.1 Introduction
- 6.2 Vector Glyphs
- 6.3 Particle Advection
- 6.4 Streamlines
- 6.5 Line Integral Convolution
- 6.6 Vector Topology
- 6.7 References

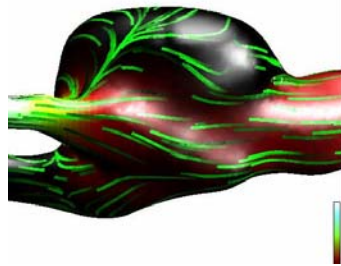
## 6.1 Introduction

- Vector fields are common in science and engineering:
  - Displacement fields in elasticity theory, velocity fields in computational fluid dynamics (CFD), force fields (e.g. gravitation), displacement fields
- In general vector fields have an orientation and are then termed *signed*, though *unsigned* vector fields, such as eigenvector fields, also exist.



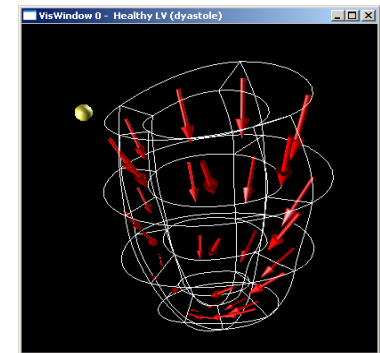
## Introduction (cont'd)

- Many visualisation techniques for vector fields were specifically developed for velocity fields.
  - *Steady flows* are constant over time.
  - *Unsteady flows* vary over time
    - Laminar flows are characterized by layers of fluid elements with similar velocities
    - Turbulent flows the velocities in neighbouring fluid elements vary randomly.



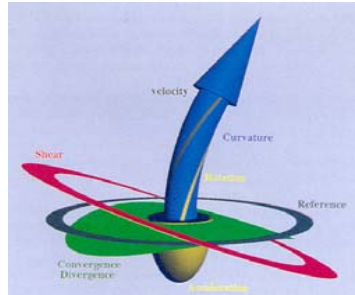
## 6.2 Vector Glyphs

- Draw arrow or line segment in the direction of the vector with length equal to the vector magnitude.
  - Advantages:
    - Good perception of visualized data (use illuminated volumetric icons for 3D vector field visualization).
  - Disadvantages:
    - Not clear which data point vector represents
    - Leads to visual cluttering
    - Requires a lot of screen space
    - Easy to miss important features



## Vector Glyphs (cont'd)

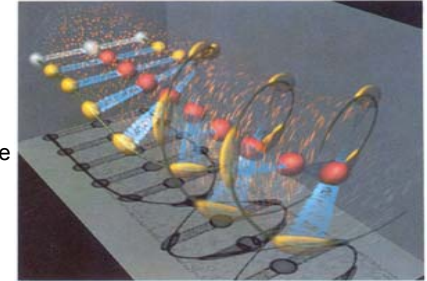
- Flow field probe (de Leeuw & van Wijk)
  - Visualises additionally neighbourhood information derived from the local velocity gradient.
    - The length, curvature and the candy stripes of the cylindrical shaft visualise magnitude, local streamline curvature and rotation of the flow field.
    - The half ellipsoid at the bottom of the shaft encodes acceleration of velocities.
    - The bending circular membrane describes convergence or divergence.
    - The angle of the ring shaped surface with respect to a reference frame encodes shear.



© 1994, Frits H. Post and Jarke J. van Wijk, Visual representation of vector fields: recent developments and research directions, in "Scientific Visualization: Advances and Challenges", Academic Press.

## 6.3 Particle Advection

- Distribute a set of particles over the domain and advect them with the vector field.
  - Flow direction and speed can be emphasized by blurring the particles.
  - Intuitive and easily understood for visualising fluid flows.
  - Well suited for turbulent flows where icons computed by integral curves and surfaces become highly irregular.
  - Lack of interactivity if the particle number is too high.
  - Difficulties in perceiving the 3D structure of the flow.

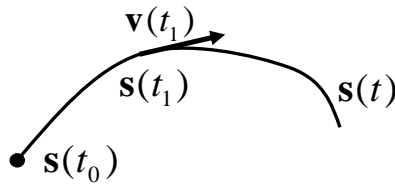


© 1994, Frits H. Post and Jarke J. van Wijk, Visual representation of vector fields: recent developments and research directions, in "Scientific Visualization: Advances and Challenges", Academic Press.

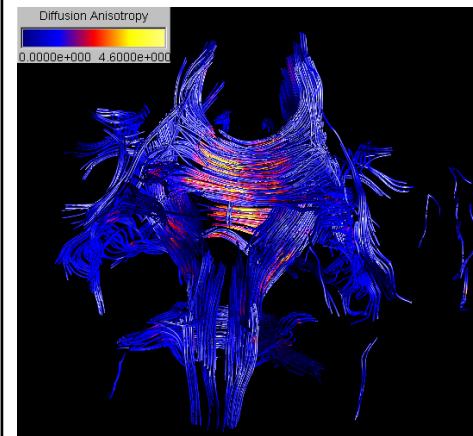
## 6.4 Streamlines

A streamline  $\mathbf{s}(t)$  for a vector field  $\mathbf{v}(t)$  is defined as the solution to the differential equation

$$\frac{\mathbf{s}(t)}{dt} = \mathbf{v}(\mathbf{s}(t)); \quad \mathbf{s}(0) = \mathbf{x}_0$$



## Streamlines (cont'd)



- Need shading and occlusion to better perceive the 3D geometry of streamlines
  - Fit thin tube around the lines

## Streamlines (cont'd)

- A simple algorithm to compute a streamline:

□ Approximate streamline by polyline  $\mathbf{x}_0 \mathbf{x}_1 \mathbf{x}_2 \mathbf{x}_3 \dots$

where  $\mathbf{x}_i = \mathbf{s}(t_i)$

$$t_0 = 0, t_{i+1} = t_i + \Delta t$$

and  $\mathbf{x}_i$  is computed by  $\mathbf{x}_{i+1} = \mathbf{x}_i + \Delta t \mathbf{v}(\mathbf{x}_i)$

## Streamlines (cont'd)

The Problem ...

- The above simple minded ODE solving method is called *Euler's Method*
- Errors accumulate steadily
- Can be unstable
  - e.g. imagine too long a time step
    - Vector at two time steps has opposite direction
    - Converging solution can blow up!
- Need to use very small step sizes to get tolerable results
  - Expensive and inefficient
- Need a better method to solve differential equations

## The mid-point method

- Can write  $\mathbf{x}_{i+1} = \mathbf{s}(t + \Delta t)$  as a Taylor expansion

$$\mathbf{s}(t + \Delta t) = \mathbf{s}(t) + \Delta t \frac{d\mathbf{s}}{dt} + \frac{\Delta t^2}{2} \frac{d^2\mathbf{s}}{dt^2} + \dots$$

- Euler's method takes first two terms on RHS. Improve by taking more.

- If take three, get  $\mathbf{s}(t + \Delta t) = \mathbf{s}(t) + \Delta t (f(\mathbf{s}(t)) + \frac{\Delta t}{2} f'(\mathbf{s}(t)))$

mid-point method:

$$\text{where } f(\mathbf{s}(t)) = \frac{d\mathbf{s}}{dt}$$

- In words:

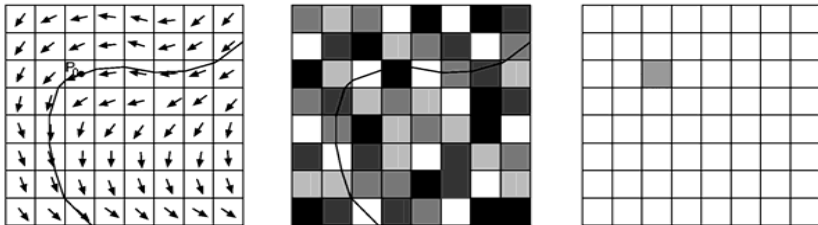
- compute Euler step to get first guess at  $\mathbf{s}(t + \Delta t)$
- Determine mid point  $\mathbf{s}(t) + \mathbf{s}(t + \Delta t)$
- Evaluate  $d\mathbf{s}/dt$  (i.e. the vector field) at this point
- Use this value to compute a new step

## Other issues

- Even mid-point method often not good enough
  - Use even higher-order methods, e.g. fourth-order Runge Kutta
- Need adaptive step sizes for best efficiency - use long time steps when things are moving slowly, short ones when changes are rapid.
- Test for vector field singularities
- Fundamental limitation of "explicit" ODE solvers:
  - Don't work well for "stiff" equations (common in computational fluid dynamics)
  - Better to use *implicit* methods

## 6.5 Line Integral Convolution

- Convolute noise texture with vector field
- Equivalent to averaging weighted pixel intensities along small streamlines



## Line Integral Convolution (cont'd)

- For any pixel  $I(q,r)$  of the input texture the centre  $\mathbf{p}_0=(q+0.5,r+0.5)$  of it is used as the centre of a streamline which is advected forwards and backwards by a length  $L$ .
- The pixels intersected by the streamline in the forward direction have the indices  $(\lfloor p_{i,x} \rfloor, \lfloor p_{i,y} \rfloor)$  where  $\mathbf{p}_i = \mathbf{p}_{i-1} + \frac{\mathbf{v}(\mathbf{p}_{i-1})}{\|\mathbf{v}(\mathbf{p}_{i-1})\|} \Delta s_{i-1}$  and  $\Delta s_{i-1}$  is the distance to the pixel boundary and  $s_0 = 0$ ,  $s_{i=1} = s_i + \Delta s_i$
- Pixels intersected in the backward direction are computed analogously and are indicated by negative indices.
- For each line segment  $[s_i, s_{i+1}]$  of the streamline intersecting pixel  $\mathbf{p}_i$  an exact integral of a convolution kernel  $k(w)$  is computed and used as weight in the LIC

$$h_i = \int_{s_j}^{s_i + \Delta s_i} k(w) dw$$

## Line Integral Convolution (cont'd)

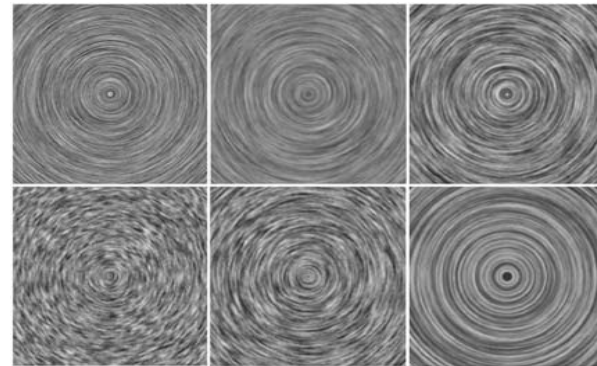
- The output pixel  $O(q,r)$  is then given by

$$O(q,r) = \frac{\sum_{i=-l}^l I(\lfloor p_{i,x} \rfloor, \lfloor p_{i,y} \rfloor) h_i}{\sum_{i=-l}^l h_i}$$

- In the simplest case the convolution kernel is a box filter so that the output texture represents the weighted input texture along the streamline.
- Vector magnitude is represented either by using colour mapping or by varying the length  $L$  of the filter kernel.

## Line Integral Convolution (cont'd)

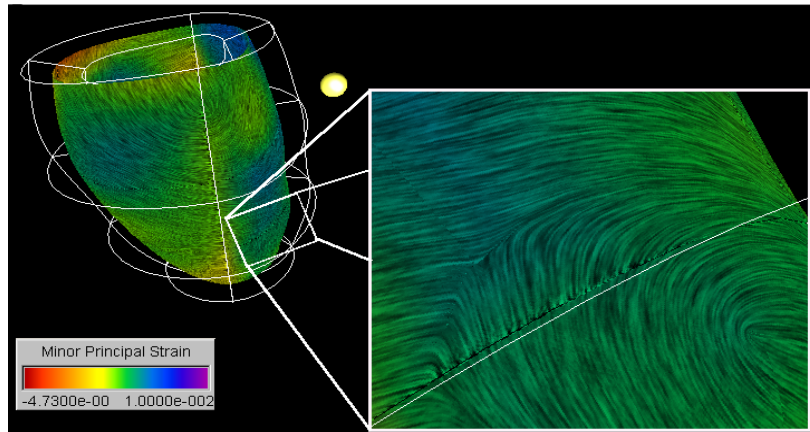
- Influence of parameters



Top row: LIC with a kernel length of 40. From left to right: using white noise, using low pass filtered white noise, using low pass filtered white noise and contrast stretching the output texture.

Bottom row: kernel length of 10, 20, and 160. All images are contrast stretched and use low-pass filtered white noise

## Line Integral Convolution (cont'd)



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## 6.6 Vector Field Topology

- A vector field  $\mathbf{v}(\mathbf{x})$  can be characterized by considering its *critical points* which are points with zero vector magnitude.
- Critical points are the only points where streamlines are non-parallel and therefore indicate important flow features.
- A critical point  $\mathbf{x}_0$  can be classified by considering the eigenvalues of the Jacobian

$$\mathbf{J}_v(\mathbf{x}_0) = \left( \frac{\partial v_i}{\partial x_j} \right) \Big|_{\mathbf{x}_0}$$

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## Vector Field Topology (cont'd)

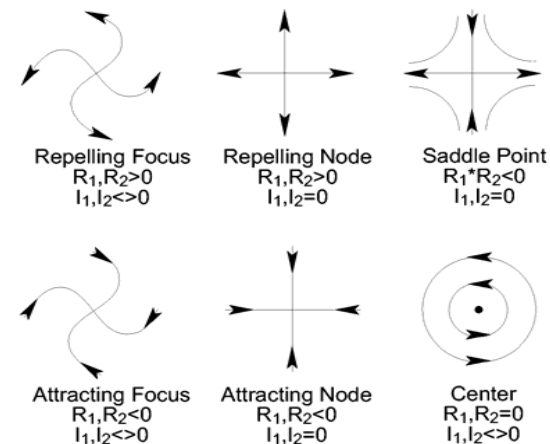
- The type of a critical point indicates the flow pattern in its immediate neighbourhood.
- In two dimensions the Jacobian of a vector field is a 2x2 matrix and therefore has two eigenvalues with real components  $R_1$  and  $R_2$  and imaginary components  $I_1$  and  $I_2$ .
- The type of a critical point and hence the local flow topology depends on the signs of these components.
  - Real components greater or smaller than zero represent repelling or attracting flow features, respectively.
  - Non-zero imaginary components symbolise circular flows.

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## Vector Field Topology (cont'd)

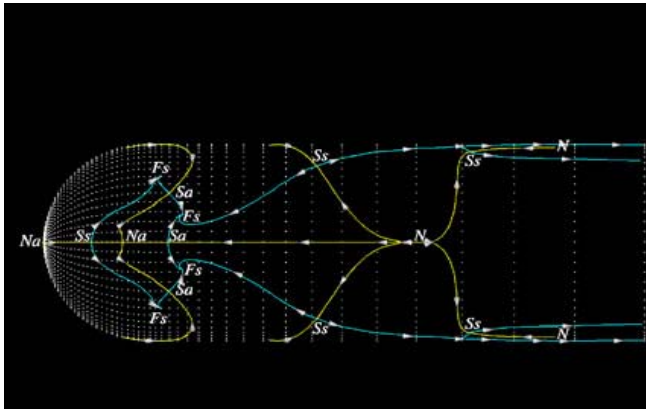


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## Vector Field Topology (cont'd)



The vector field topology is obtained by connecting critical points by special streamlines

© 1994, Lambertus Hesselink and Thierry Delmarcelle, Chapter 26: Visualization of vector and tensor data sets, in "Scientific Visualization: Advances and Challenges", Academic Press.

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