

Introduction to Scientific Visualization

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Visualization – Definition

visualization: to form a **mental vision, image,** or **picture** of (something not visible or present to the sight, or of an abstraction); to make **visible to the mind or imagination**

[Oxford Engl. Dict., 1989]



tool to enable a user insight into data

“The purpose of computing is **insight**, not numbers.”

[R. Hamming, 1962]

Visualization – Goals

■ Visualization, ...

- ◆ ... to **explore**
 - Nothing is known,
Vis used for **data exploration**
- ◆ ... to **analyze**
 - There are hypotheses, **verification or falsification**
- ◆ ... to **present**
 - “everything” known about the data, **communication of results**

Visualization – Areas

■ Three major areas

- ◆ Volume Visualization
 - ◆ Flow Visualization
- } Scientific Visualization
-
- ◆ Information Visualization
- usually no spatial reference
- inherent spatial reference
- 2D/3D
- nD

InfoVis vs. SciVis

■ N-dimensional vs. 2/3-dimensional

- ◆ SciVis can be N-dimensional too (time series, simulation data, ...)

■ Abstract data vs. spatial data

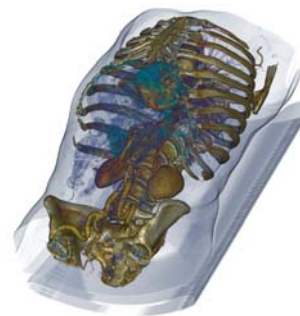
- ◆ InfoVis data may also have spatial attributes (country, state, ...)

■ Discrete data vs. continuous data

- ◆ InfoVis data may be sampled from a continuous domain

SciVis – Examples (1)

■ Volume data



SciVis – Examples (2)

- Flow data

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Medicine

sketch from Leonardo Da Vinci's anatomical notebooks

medical illustrations by Clarice Ashworth Francone

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Cartography

isolines to visualize compass deviations

wind flow visualization

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Meteorology

map with iso-pressure lines

weather fronts

map for pilots

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Experimental Flow Investigation

- Fixation of tufts, ribbons on ...
 - aircraft in wind tunnels
 - ship hull in fluid tanks
- Introduction of smoke particles (in wind tunnel)
- Introduction of dye (in fluids)

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Visualization Scenarios

complexity, tech. demands

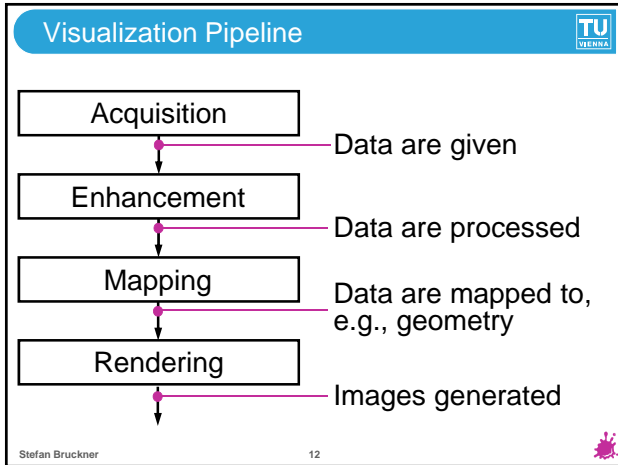
benefits, possibilities

passive visualization

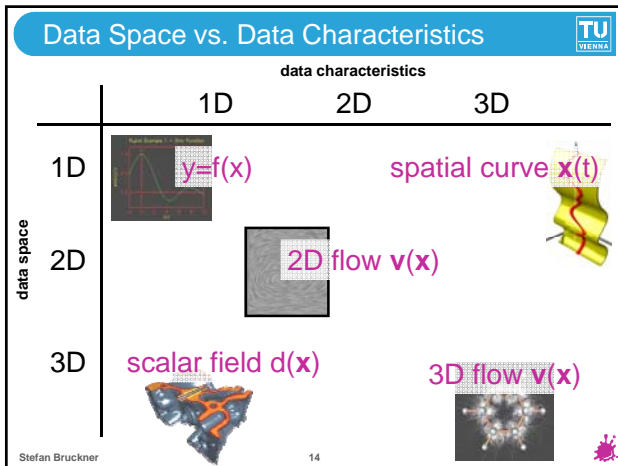
interactive visualization

interactive steering

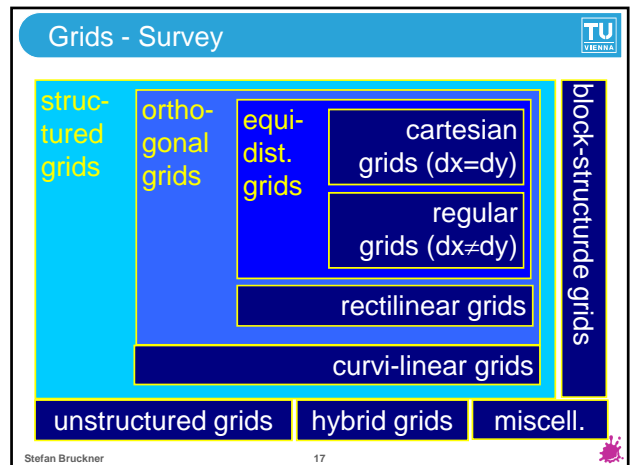
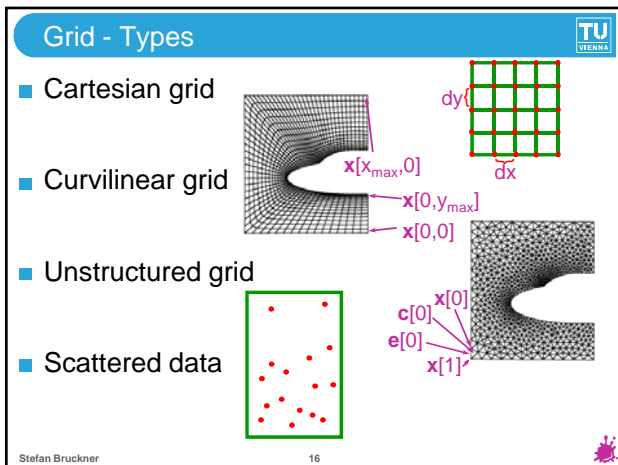
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- ### Data
- Focus of visualization, everything is centered around the data
 - Driving factor (besides user) in choice and attribution of the visualization technique
 - Important questions
 - ◆ In what domain are the data given? (**data space**)
 - ◆ What is the type of data? (**data characteristics**)
 - ◆ Which **representation** makes sense?
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- ### Grids – General Information
- Important questions
 - ◆ Which data organization is optimal?
 - ◆ Where do the data come from?
 - ◆ Is there a neighborhood relationship?
 - ◆ How is the neighborhood information stored?
 - ◆ How is navigation within the data possible?
 - ◆ Calculations with the data possible?
 - ◆ Are the data structured?
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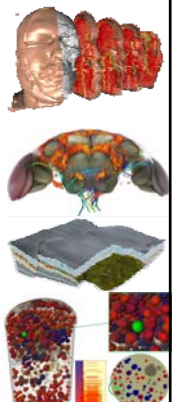
Volume Visualization

- **VolVis = visualization of volume data**
 - ◆ Mapping 3D→2D
- **Volume data**
 - ◆ 3D×1D data
 - ◆ Scalar data, 3D data space, space filling
- **User goals**
 - ◆ Gain insight in 3D data
 - ◆ Structures of special interest + context

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Volume Data

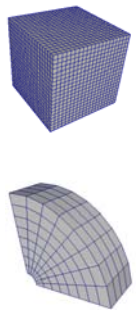
- **Medicine**
 - ◆ CT, MRI, PET, Ultrasound
- **Biology**
 - ◆ Confocal microscopy, histological cuts
- **Geology**
 - ◆ Seismic surveys
- **Material testing**
 - ◆ Industrial CT



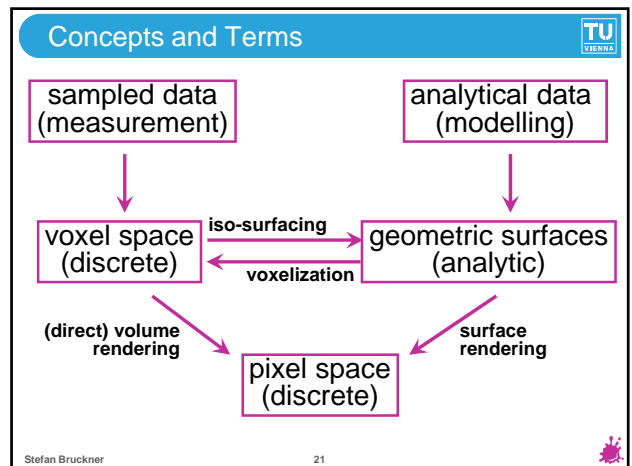
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3D Data Space


- **Cartesian/regular grid**
 - ◆ Most common, e.g., CT/MRI scans
- **Curvilinear/unstructured grid**
 - ◆ Less frequently, e.g., simulation data



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Volume Rendering (1)



- Deals with the visual representation of 3D functions
- Frequently, but not exclusively, functions are scalar-valued
- Often acquired using sampling (e.g., medical domain)

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Volume Rendering (2)

- Initially volumes were visualized using two-dimensional cuts
- Extraction of surface geometry for isosurfaces in the volume (e.g. Marching Cubes [Lorensen and Cline 1987])
- Volume rendering introduced almost simultaneously by [Levoy 1988] and [Drebin et al. 1988]

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Surface vs. Volume Rendering



- Surface rendering
 - ◆ **Indirect** volume visualization
 - ◆ Intermediate representation: iso-surface
 - ◆ Pros: Less memory, fast rendering
- Volume rendering
 - ◆ **Direct** volume visualization
 - ◆ Usage of transfer functions
 - ◆ Pros: illustrate the interior, semi-transparency

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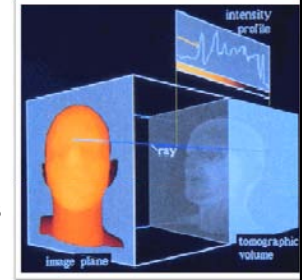
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Volume Ray Casting



- **Volume:** 1D value defined in 3D $f(\mathbf{x}) \in \mathbb{R}^1, \mathbf{x} \in \mathbb{R}^3$
- **Ray:** Half-line $\mathbf{r}(t) \in \mathbb{R}^3, t \in \mathbb{R}^1 > 0$
- **Intensity profile:** values along a ray $f(\mathbf{r}(t)) \in \mathbb{R}^1, t \in \mathbb{R}^1 > 0$
- **Image plane:** starting points of rays



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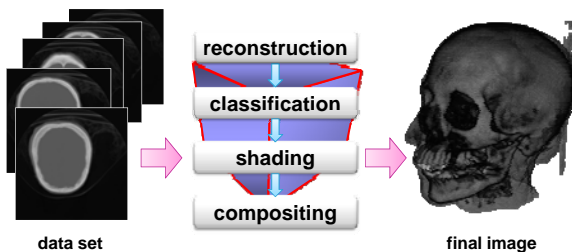
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Pipeline – Overview



volume rendering pipeline



data set

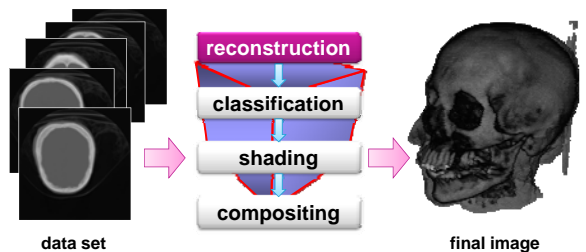
final image



Pipeline – Reconstruction



volume rendering pipeline



data set

final image



Reconstruction (1)



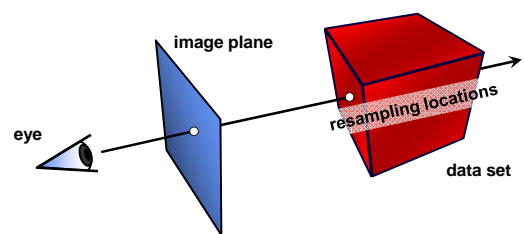
- Usually volume data sets are given as a grid of discrete samples
- For rendering purposes, we want to treat them as continuous three-dimensional functions
- We need to choose an appropriate reconstruction filter
- Requirements: high-quality reconstruction, but small performance overhead

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Reconstruction (2)



eye

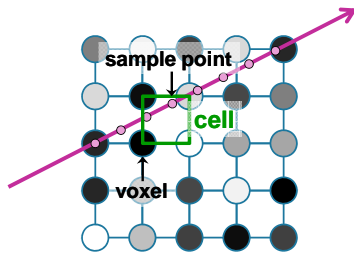
image plane

resampling locations

data set



Reconstruction (3)



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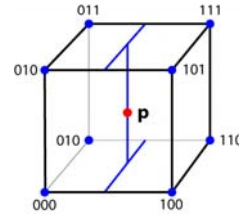
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Trilinear Interpolation



- Simple extension of linear interpolation to three dimensions
- Advantage: current GPUs automatically do trilinear interpolation of 3D textures



$$v_p = v_{000}(1-x_p)(1-y_p)(1-z_p) + v_{100}x_p(1-y_p)(1-z_p) + v_{010}(1-x_p)y_p(1-z_p) + v_{001}(1-x_p)(1-y_p)z_p + v_{011}(1-x_p)y_pz_p + v_{101}x_p(1-y_p)z_p + v_{110}x_py_p(1-z_p) + v_{111}x_py_pz_p$$

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Other Reconstruction Filters



- If very high quality is required, more complex reconstruction filters may be required
- Marschner-Lobb function is a common test signal to evaluate the quality of reconstruction filters [Marschner and Lobb 1994]
- The signal has a high amount of its energy near its Nyquist frequency
- Makes it a very demanding test for accurate reconstruction

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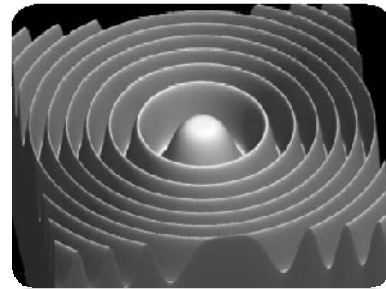
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Comparison of Reconstruction Filters (1)



- Marschner-Lobb test signal (analytically evaluated)



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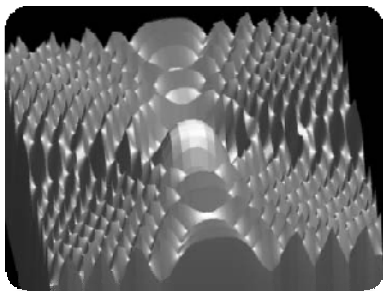
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Comparison of Reconstruction Filters (2)



- **Trilinear** reconstruction of Marschner-Lobb test signal



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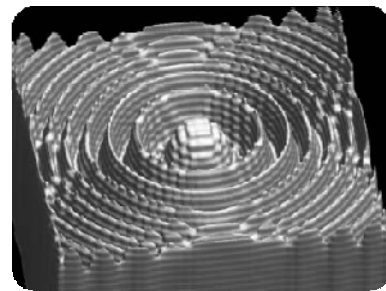
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Comparison of Reconstruction Filters (3)



- **Cubic** reconstruction of Marschner-Lobb test signal



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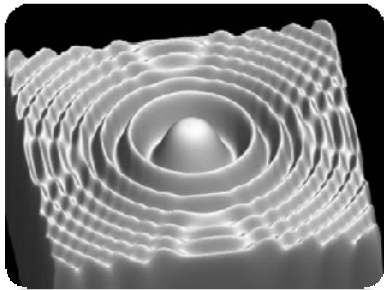
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Comparison of Reconstruction Filters (4)



- **B-Spline** reconstruction of Marschner-Lobb test signal



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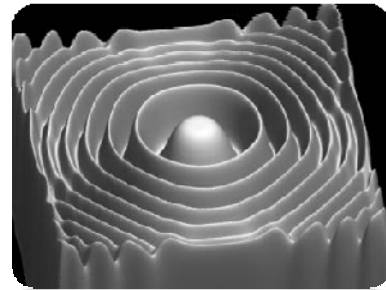
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Comparison of Reconstruction Filters (5)



- **Windowed sinc** reconstruction of Marschner-Lobb test signal



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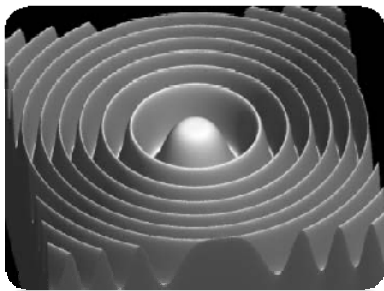
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Comparison of Reconstruction Filters (6)



- Marschner-Lobb test signal (analytically evaluated)



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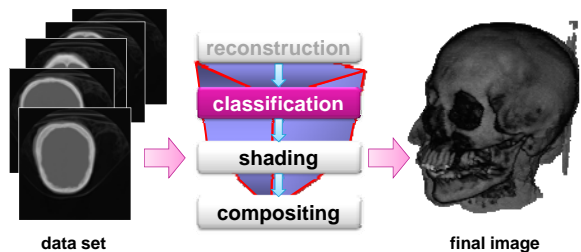
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Pipeline – Classification



volume rendering pipeline



Classification (1)



- Projecting a 3D data set onto a 2D image is problematic
- Not all information contained in the volume is relevant to the user
- Classification allows the user to extract the important parts of the data

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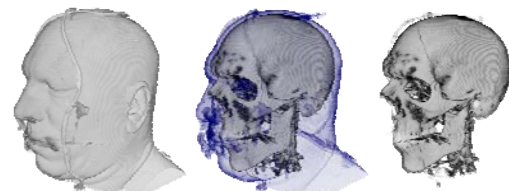
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Classification (2)



- During Classification the user defines the appearance of the data
 - ◆ Which parts are transparent?
 - ◆ Which parts have which color?



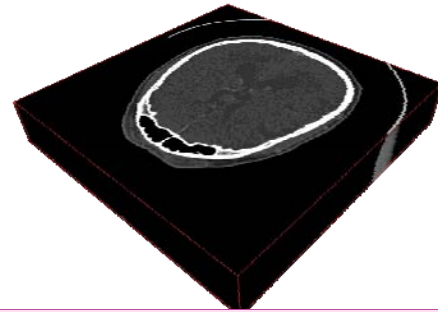
Classification (3)



- During Classification the user defines the appearance of the data
 - ◆ Which parts are transparent?
 - ◆ Which parts have which color?
- The user defines a **transfer function**



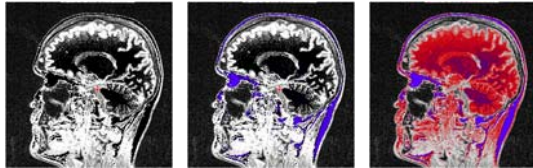
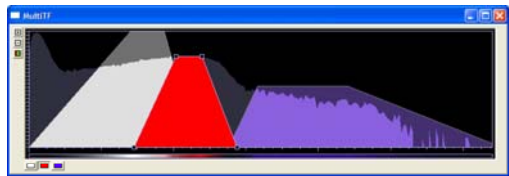
Transfer Functions (1)



real-time update of the transfer function important



Transfer Functions (2)



Classification Order (1)



- Classification can occur before or after reconstruction
- **Pre-interpolative:** classify all data values and then interpolate between RGBA-tuples
- **Post-interpolative:** interpolate between scalar data values and then classify the result

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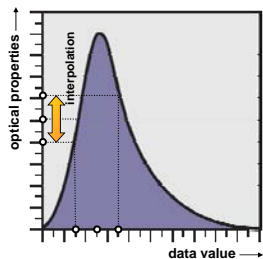
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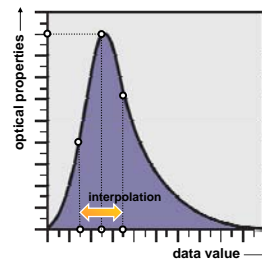
Classification Order (2)



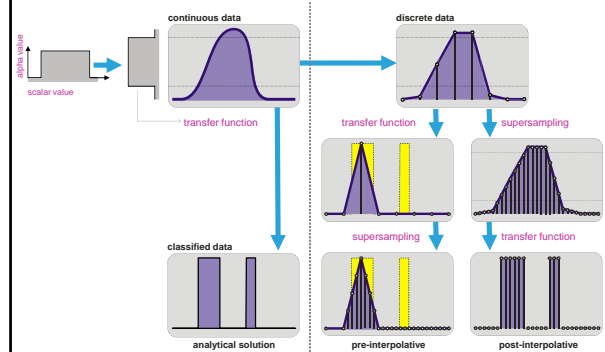
PRE-INTERPOLATIVE



POST-INTERPOLATIVE



Classification Order (3)



Classification Order (4)

pre-interpolative post-interpolative

same transfer function, resolution, and sampling rate

Classification Order (5)

pre-interpolative post-interpolative

same transfer function, resolution, and sampling rate

Pipeline – Shading

volume rendering pipeline

data set → reconstruction → classification → shading → compositing → final image

Shading (1)

- Make structures in volume data sets more realistic by applying an illumination model
- Shade each sample in the volume like a surface
- Any model used in real-time surface graphics suitable
- Common choice: Blinn-Phong illumination model

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Shading (2)

- Local illumination, similar to surface lighting
 - ◆ Lambertian reflection
light is reflected equally in all directions
 - ◆ Specular reflection
light is reflected scattered around the direction of perfect reflection

Shading (3)

shaded volume rendering unshaded volume rendering

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Gradient Estimation (1)



- Normalized gradient vector of the scalar field is used to substitute for the surface normal
- The gradient vector is the first-order derivative of the scalar field

$$\nabla f(\mathbf{x}) = \begin{pmatrix} \frac{\partial f(\mathbf{x})}{\partial x} \\ \frac{\partial f(\mathbf{x})}{\partial y} \\ \frac{\partial f(\mathbf{x})}{\partial z} \end{pmatrix}$$

partial derivative in x-direction

partial derivative in y-direction

partial derivative in z-direction



Gradient Estimation (2)



- We can estimate the gradient vector using finite differencing schemes, e.g. central differences:

$$\nabla f(x, y, z) \approx \frac{1}{2h} \begin{pmatrix} f(x+h, y, z) - f(x-h, y, z) \\ f(x, y+h, z) - f(x, y-h, z) \\ f(x, y, z+h) - f(x, y, z-h) \end{pmatrix}$$

- Noisy data may require more complex estimation schemes



Gradient Magnitude



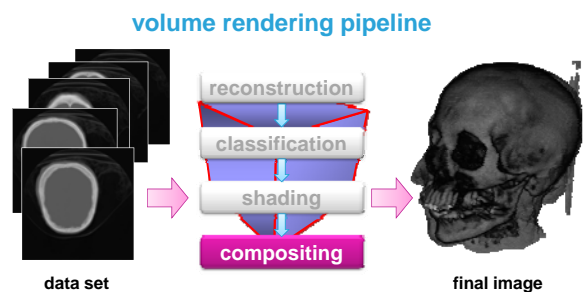
- Magnitude of gradient vector can be used to measure the “surfacedness” of a point
 - Strong changes → high gradient magnitude
 - Homogeneity → low gradient magnitude
- Applications
 - Use gradient magnitude to modulate opacity of sample
 - Interpolate between unshaded and shaded sample color using gradient magnitude as weight

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Pipeline – Compositing



Compositing (1)



- So far, everything discussed applies to single sample points along a viewing ray
- How to subsequent sample when traversing the ray?
- Common models
 - Maximum Intensity Projection
 - Emission-Absorption Model

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Maximum Intensity Projection (1)



- Always display the maximum value along a viewing ray
- Motivation: visualization of contrast-enhanced tomographic scans
- Parameterless rendering, very common in medical domain

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Maximum Intensity Projection (2)



- Problem: loss of spatial relationships between different structures



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Emission-Absorption Model (1)



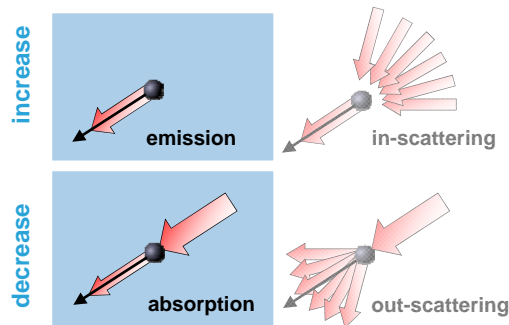
- Conventional volume rendering uses an emission-absorption model
- Scattering effects are usually ignored due to high computational complexity
- For each pixel on the image plane, a the ray integral has to be solved
- For each step along the viewing ray, perform accumulate RGBA from transfer function

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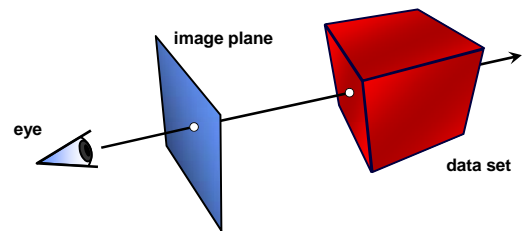
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Emission-Absorption Model (2)



Ray Integration (1)

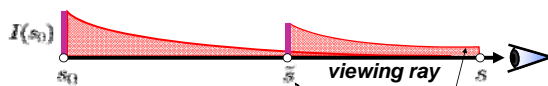


Ray Integration (2)



How do we determine the radiant energy along the ray?

Physical model: emission and absorption, no scattering



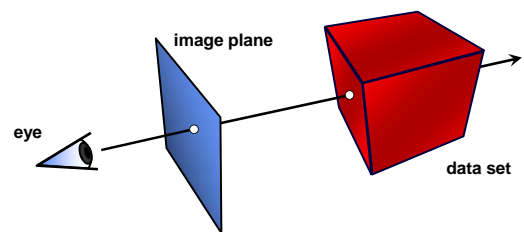
Every point \bar{s} along the viewing ray emits additional radiant energy

active emission at point \bar{s} absorption along the distance $s - \bar{s}$

$$I(s) = I(s_0) e^{-\tau(s_0, s)} + \int_{s_0}^s q(\bar{s}) e^{-\tau(\bar{s}, s)} d\bar{s}$$



Numerical Solution (1)



Numerical Solution (2)

$$\bar{C} = \sum_{i=0}^{[T/\Delta t]} C_i \prod_{j=0}^{i-1} (1 - A_j)$$

can be computed recursively

$$C_i = C_i + (1 - A_i)C_{i-1}$$

radiant energy
observed at position i

radiant energy
emitted at position i

absorption at
position i

radiant energy
observed at position $i-1$

Numerical Solution (2)

back-to-front compositing

$$C_i = C_i + (1 - A_i)C_{i-1}$$

front-to-back compositing

$$C_i = C_{i+1} + (1 - A'_{i+1})C_i$$

$$A'_i = A'_{i+1} + (1 - A'_{i+1})A_i$$

early ray termination:
stop the calculation
when
 $A'_i \approx 1$

Further Reading

- M. Levoy. Display of Surfaces from Volume Data. *IEEE Computer Graphics and Applications*, 8(3):29-37, 1988.
- R. Drebin, L. Carpenter, P. Hanrahan. Volume Rendering. *ACM SIGGRAPH Computer Graphics*, 22(4):65-74, 1988.
- W. Lorensen, H. Cline. Marching cubes: A high resolution 3D surface construction algorithm. *ACM SIGGRAPH Computer Graphics*, 21(4):163-169, 1987.
- C. Rezk-Salama, K. Engel, M. Hadwiger, J. Kniss, D. Weiskopf. *Real-Time Volume Graphics*, AK Peters, 1-56881-266-3, 2006.

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Flow Visualization

- **FlowVis = visualization of flows**
 - ◆ Visualization of change information
- **Flow data**
 - ◆ nD×nD data, 1D²/2D²/nD² (models), 2D²/3D²
 - ◆ Vector data (nD) in nD data space
 - ◆ Steady vs. time-dependent flow
- **User goals**
 - ◆ Overview vs. details (with context)

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Flow Data (1)


- **Simulation**
 - ◆ Flow space modelled with grid
 - ◆ FEM (finite elements method), CfD (computational fluid dynamics)
- **Measurements**
 - ◆ Optical methods + pattern recognition, e.g.: PIV (particle image velocimetry)
- **Models**
 - ◆ Differential equation systems dx/dt

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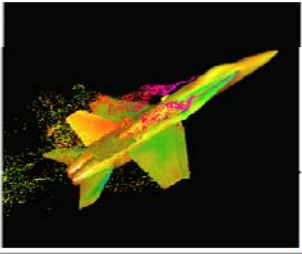
Flow Data (2)

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Flow Data (3)



experiment

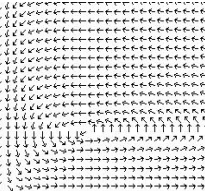
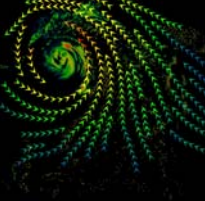


simulation

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Direct Flow Visualization

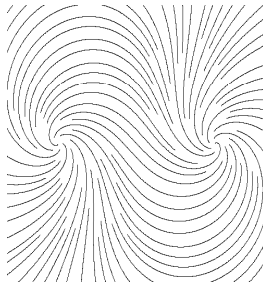

- Grid of arrows to visualize flow directions
- Normalized arrows vs. scaling with velocity
- Quite effective in 2D, problematic in 3D
- Sometimes limited expressivity (temporal component missing)

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Geometric Flow Visualization

- Idea: follow the flow in time (integration) to extract the path of a particle

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Streamlines – Theory

- Flow data \mathbf{v} : derivative information
 - $\frac{d\mathbf{x}}{dt} = \mathbf{v}(\mathbf{x})$
spatial points $\mathbf{x} \in \mathbb{R}^n$, flow vectors $\mathbf{v} \in \mathbb{R}^n$, time $t \in \mathbb{R}$
- Streamline \mathbf{s} : integration over time, also called trajectory, solution, curve
 - $\mathbf{s}(t) = \mathbf{s}_0 + \int_{0 \leq u \leq t} \mathbf{v}(\mathbf{s}(u)) du$
seed point \mathbf{s}_0 , integration variable u
- Difficulty: result \mathbf{s} also in the integral, analytical solution usually impossible

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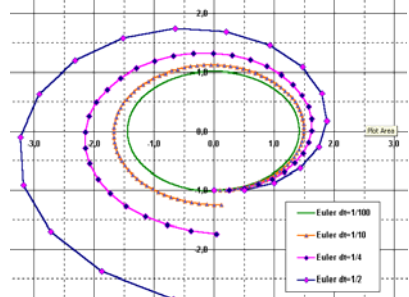
Streamlines – Practice (1)

- Solve using numerical integration techniques
- Assume that locally the solution is approximately linear
- Euler integration
 - $\mathbf{s}_{i+1} = \mathbf{s}_i + dt \cdot \mathbf{v}(\mathbf{s}_i)$
Follow the current flow vector $\mathbf{v}(\mathbf{s}_i)$ from the current streamline point \mathbf{s}_i for a short time (dt)

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Streamlines – Practice (2)

- Accuracy of results is strongly dependent on step size dt

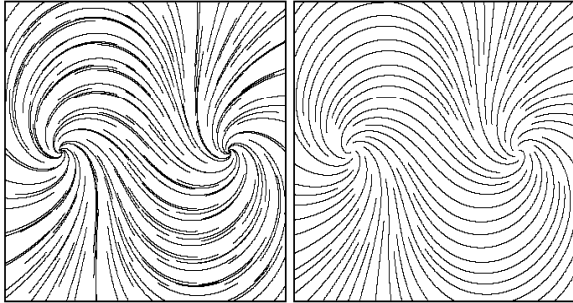


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Streamlines – Placement (1)



- Seed fill with streamlines to achieve equal density



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Streamlines – Placement (2)

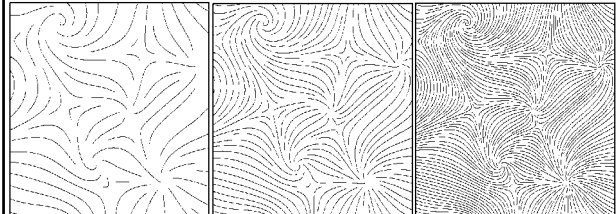


- Variations of distance in relation to image width

6%

3%

1.5%



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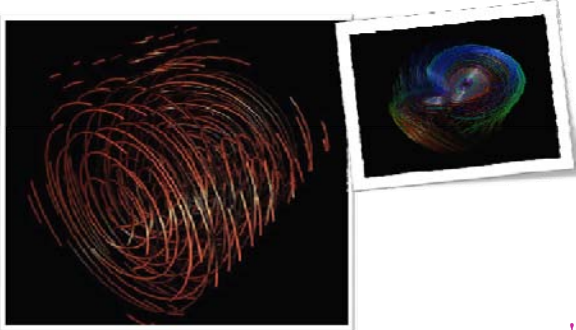
79



Illuminated Streamlines



- Illuminated 3D curves improve perception



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Stream Surfaces



- Natural extension of streamlines to 3D
- Surfaces which are tangential to the vector field everywhere
- Challenges related to occlusion and visual complexity



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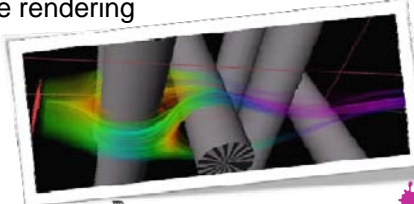
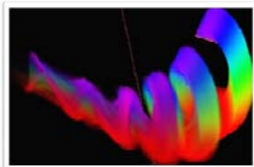
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Flow Volumes



- Volumetric equivalents of streamlines, subset of a 3D flow domain is traced in time
- Can be visualized with direct volume rendering methods



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Unsteady Flow



- Path line
 - ◆ Trajectory of an individual particle in the fluid flow
- Timeline
 - ◆ Joins the positions of particles released at the same instant in time
- Streak line
 - ◆ Connects particles that have passed through a certain point in space

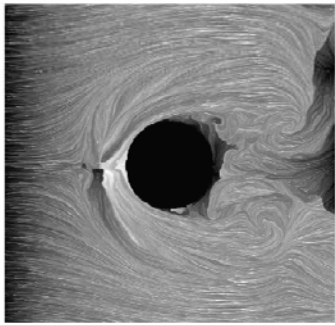
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Texture-based Flow Visualization

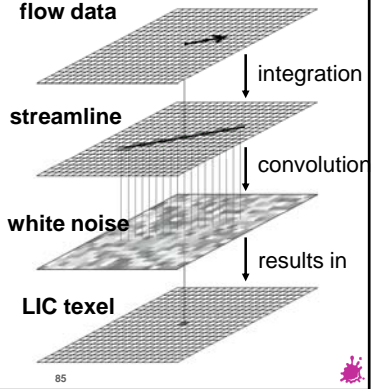
- Idea: exploit visual correlations to provide a dense visualization of the flow



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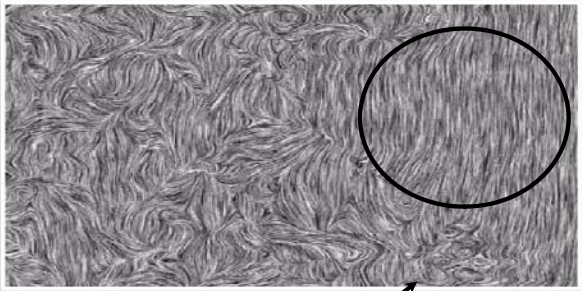
Line Integral Convolution

- Calculation of a texture value
 - look at streamline through point
 - filter white noise along streamline



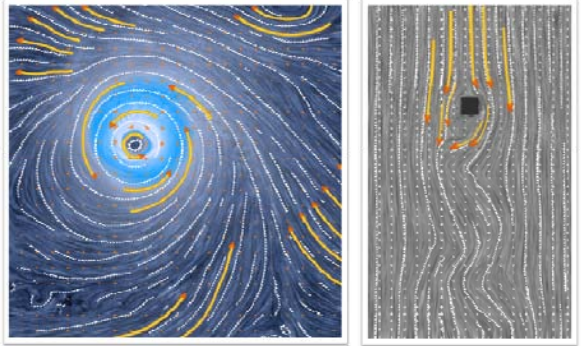
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Line Integral Convolution – Examples (1)



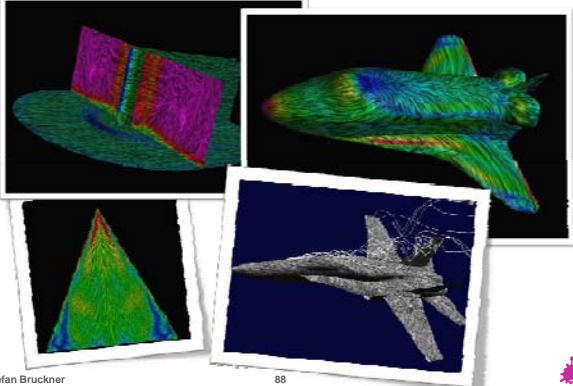
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Line Integral Convolution – Examples (2)



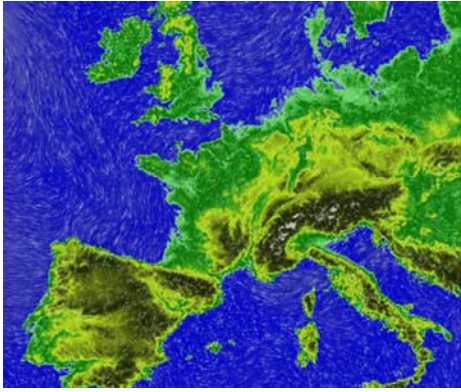
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Line Integral Convolution on Surfaces



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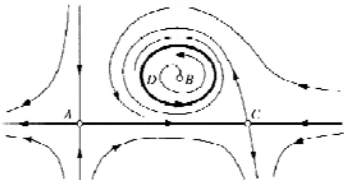
Texture Advection – Unsteady Flows



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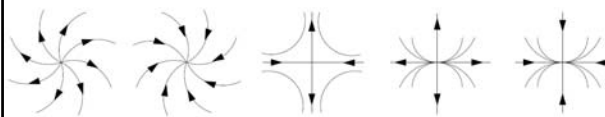
Feature-based Flow Visualization

- Extract and visualize the abstract structure of a flow
- Different elements
 - Checkpoints, defined through $v(x)=0$
 - Cycles, defined through $sx(t+T)=sx(t)$
 - Connecting structures (separatrices, etc.)



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Vector Field Topology

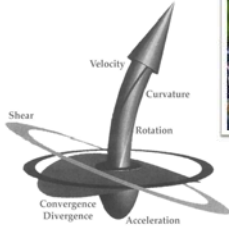

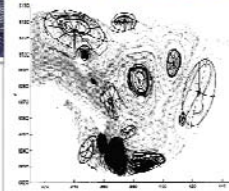


- Critical points can be classified by the Eigenvalues of the Jacobian
 R = real components, I = imaginary components

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Glyphs/Icons


- Local/topological properties

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Examples (1)

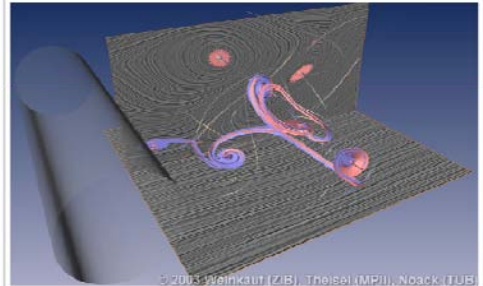
- Topology of a hurricane simulation



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Examples (2)

- Visualization of flow past a circular cylinder using critical points and saddle connectors



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Further Reading

- R. Laramée, H. Hauser, H. Doleisch, B. Vrolijk, F. Post, D. Weiskopf. The State of the Art in Flow Visualization: Dense and Texture-Based Techniques. *Computer Graphics Forum*, 23(2):203-221, 2004.
- F. Post, B. Vrolijk, H. Hauser, R. Laramée, H. Doleisch. The State of the Art in Flow Visualization: Feature Extraction and Tracking. *Computer Graphics Forum*, 22(4):775-792, 2003.

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Summary



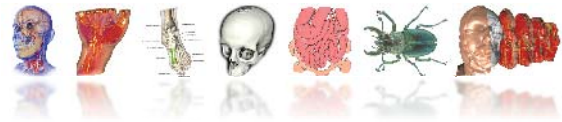
- Scientific visualization is data-driven, but it is crucial to keep the goal of the user in mind
- **Volume visualization**
3D scalar data
 - ◆ Important to provide detailed view of structures of interest
- **Flow visualization**
2D/3D vector data
 - ◆ Provide overview and characterize flow behavior

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Thank you for your attention!



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Meister Eduard Gröller
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