

# Volume Visualization

## Part 1 (out of 3)

### Volume Data

- Where do the data come from?
  - ◆ **Medical Application**
    - Computed Tomographie (CT)
    - Magnetic Resonance Imaging (MR)
  - ◆ **Materials testing**
    - Industrial-CT
  - ◆ **Simulation**
    - Finite element methods (FEM)
    - Computational fluid dynamics (CFD)
  - ◆ **etc.**

Eduard Gröller, Helwig Hauser

4

### Overview: Volume Visualization

- Introduction to volume visualization
  - ◆ On volume data
  - ◆ Surface vs. volume rendering
  - ◆ Overview: Techniques
- Simple methods
  - ◆ Slicing, cuberille
- Direct volume visualization
  - ◆ Introduction, types of combinations
  - ◆ Transfer functions

Eduard Gröller, Helwig Hauser

2

### 3D Data Space

- How are volume data organized?
  - ◆ **Cartesian resp. regular grid:**
    - CT/MR: often  $dx=dy<dz$ , e.g. 135 slices (z) á 512<sup>2</sup> values (as x & y pixels in a slice)
    - **Data enhancement:** iso-stack-calculation = Interpolation of additional slices, so that  $dx=dy=dz \Rightarrow 512^3$  Voxel
    - Data: **Cells** (cuboid), Corner: **Voxel**
  - ◆ **Curvi-linear grid resp. unstructured:**
    - Data organized as tetrahedra or hexahedra
    - Often: conversion to tetrahedra

Eduard Gröller, Helwig Hauser

5

### Volume Visualization

- Introduction:
  - ◆ **VolVis = visualization of volume data**
    - Mapping 3D→2D
    - Projection (e.g., MIP), slicing, vol. rendering, ...
  - ◆ **Volume data =**
    - 3D×1D data
    - Scalar data, 3D data space, space filling
  - ◆ **User goals:**
    - Gain insight in 3D data
    - Structures of special interest + context

Eduard Gröller, Helwig Hauser

3

### VolVis – Challenges

- **Rendering projection,** so much information and so few pixels!
- **Large data sizes,** e.g. 512×512×1024 voxel á 16 bit = 512 Mbytes
- **Speed,** Interaction is very important, >10 fps!

Eduard Gröller, Helwig Hauser

6

### Voxels vs. Cells

- Two ways to interpret the data:
  - Data: set of voxel
    - voxel** = abbreviation for volume element (cf. pixel = "picture elem.")
    - voxel = point sample in 3D
    - Not necessarily interpolated
  - Data: set of cells
    - cell = cube primitive (3D)
    - Corners: 8 voxel (see above)
    - Values in cell: interpolation used

Eduard Gröller, Helwig Hauser 7

### Gradients as Normal Vector Replacement

- Gradient  $\nabla f = (\partial f/\partial x, \partial f/\partial y, \partial f/\partial z)$
- $\nabla f|_{x_0}$  normal vector to iso-surface  $f(x_0)=f_0$
- Central difference in x-, y- & z-direction (in voxel):
 
$$\nabla f(x,y,z) = 1/2 \begin{pmatrix} f(x+1)-f(x-1) \\ f(y+1)-f(y-1) \\ f(z+1)-f(z-1) \end{pmatrix}$$
- Then tri-linear interpolation within a cell
- Alternatives:
  - Forward differencing:  $\nabla f(x)=f(x+1)-f(x)$
  - Backwards differencing:  $\nabla f(x)=f(x)-f(x-1)$
  - Intermediate differencing:  $\nabla f(x+0.5)=f(x+1)-f(x)$

Eduard Gröller, Helwig Hauser 10

### Interpolation

$$v = (1-x)(1-y)(1-z)S(0,0,0) + (x)(1-y)(1-z)S(1,0,0) + (1-x)(y)(1-z)S(0,1,0) + (x)(y)(1-z)S(1,1,0) + (1-x)(1-y)(z)S(0,0,1) + (x)(1-y)(z)S(1,0,1) + (1-x)(y)(z)S(0,1,1) + (x)(y)(z)S(1,1,1)$$

$v = S(\text{rnd}(x), \text{rnd}(y), \text{rnd}(z))$

Nearest Neighbor                      Trilinear

Eduard Gröller, Helwig Hauser 11

### Concepts and Terms

Eduard Gröller, Helwig Hauser 11

### Interpolation – Results

Nearest Neighbor Interpolation                      Trilinear Interpolation

Eduard Gröller, Helwig Hauser

### Concepts and Terms

- Example 1:
  - CT measurement
  - Iso-stack-conversion
  - Iso-surface-calculation (marching cubes)
  - Surface rendering (OpenGL)

Eduard Gröller, Helwig Hauser

### Concepts and Terms

- Example 2:
  - MR measurement
  - Iso-stack-conversion
  - MIP (maximum intensity proj.)
  - Image: blood-vessels in hand

The flowchart illustrates the pipeline for Example 2. It starts with 'sampled data (measurement)' which is converted to 'voxel space (discrete)' via 'iso-surfacing'. Simultaneously, 'analytic data (modelling)' is converted to 'geom. surfaces (analytic)' via 'voxelization'. 'voxel space (discrete)' is then rendered to 'pixel space (discrete)' through 'volume rendering', which produces the final image of blood vessels. 'geom. surfaces (analytic)' are rendered to 'pixel space (discrete)' through 'surface rendering'.

Eduard Gröller, Helwig Hauser

### Surfaces vs. Volume Rendering

- Surface rendering:
  - Indirect volume visualization
  - Intermediate representation: iso-surface, "3D"
  - Pros: Shading → Shape!, HW-rendering
- Volume rendering:
  - Direct volume visualization
  - Usage of transfer functions
  - Pros: illustrate the interior, semi-transparency

Eduard Gröller, Helwig Hauser

### Concepts and Terms

- Example 3:
  - potential function  $\rho(x,y,z)$
  - Iso-surface  $\rho(x,y,z) = \rho_0$
  - Surface: ray tracing

The flowchart for Example 3 shows 'sampled data (measurement)' leading to 'voxel space (discrete)' via 'iso-surfacing'. 'analytic data (modelling)' leads to 'geom. surfaces (analytic)' via 'voxelization'. 'voxel space (discrete)' is rendered to 'pixel space (discrete)' through 'volume rendering', resulting in a red surface image. 'geom. surfaces (analytic)' are rendered to 'pixel space (discrete)' through 'surface rendering'.

Eduard Gröller, Helwig Hauser

### Surfaces vs. Volume Rendering

This slide compares 'hybrid rendering = surfaces + volumes' and 'volume rendering'. It shows a sequence of images for a star-shaped object. The top row shows the object as a series of surfaces. The bottom row shows the object as a volume with internal structure. The hybrid rendering combines both, showing the surface and internal volume together.

Eduard Gröller, Helwig Hauser

### Concepts and Terms

- Example 4:
  - X-Ray Modelling
  - Surface-definition
  - Sampling (voxelization), combination
  - Direct volume rendering

The flowchart for Example 4 shows 'sampled data (measurement)' leading to 'voxel space (discrete)' via 'iso-surfacing'. 'analytic data (modelling)' leads to 'geom. surfaces (analytic)' via 'voxelization'. 'voxel space (discrete)' is rendered to 'pixel space (discrete)' through 'volume rendering', resulting in a brain scan image. 'geom. surfaces (analytic)' are rendered to 'pixel space (discrete)' through 'surface rendering'.

Eduard Gröller, Helwig Hauser

### VolVis-Techniques – Overview

- Simple methods:
  - Slicing, MPR (multi-planar reconstruction)
- Direct volume visualization:
  - Ray casting
  - Shear-warp factorization
  - Splatting
  - 3D texture mapping
  - Fourier volume rendering
- Surface-fitting methods:
  - Marching cubes (marching tetrahedra)

Eduard Gröller, Helwig Hauser

## Image-Order vs. Object-Order



- Image-order:
  - ◆ FOR every pixel DO: ...
  - ◆ Cost, complexity  $\approx$  image size
  - ◆ Example: ray casting (tracing viewing rays)
- Object-order:
  - ◆ FOR every object (voxel) DO: ...
  - ◆ Cost, complexity  $\approx$  object size (# of voxels)
  - ◆ Examples: splatting ("throwing snow balls")



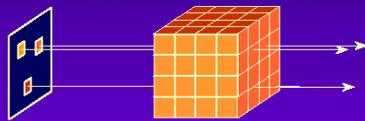
## Simple Methods

Slicing, etc.

## Image-Order Approach



Image-Order Approach: Traverse the image pixel-by-pixel and sample the volume.

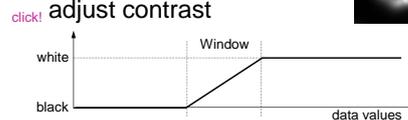
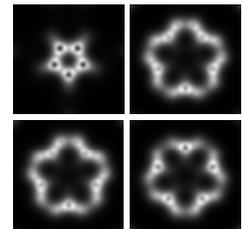


Ray Casting

## Slicing



- Slicing:
  - ◆ Axes-parallel slices
  - ◆ regular grids: simple
  - ◆ without transfer function no color
  - ◆ Windowing: adjust contrast



## Object-order approach



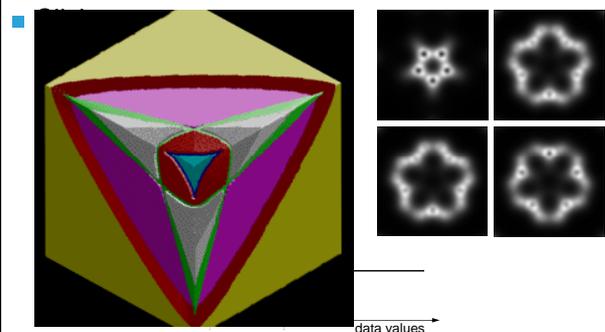
Object-Order Approach: Traverse the volume, and project to the image plane.



Splatting  
cell-by-cell

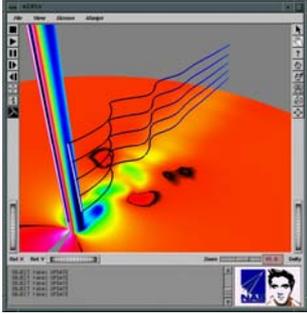
Texture Mapping  
plane-by-plane

## Slicing



## Slicing

- Not so simple:
  - ◆ Slicing through general grid
  - ◆ Interpolation necessary
- Slicing:
  - ◆ well combinable with 3D-visualization
- Multi-planar reformation (MPR):
  - ◆ arbitrary axes, 3D



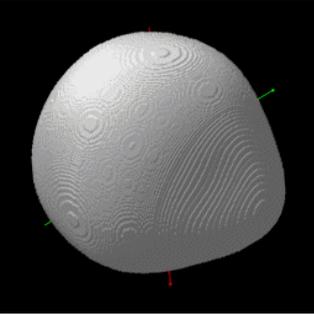
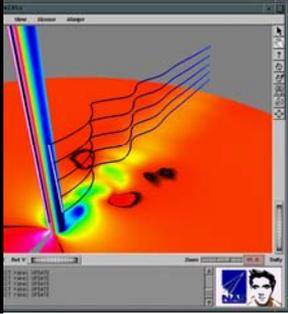
Eduard Gröller, Helwig Hauser 25

## Direct Volume Visualization

- Overview:
  - ◆ No intermediate representation
  - ◆ “real 3D”
  - ◆ Integration of so much information difficult
  - ◆ Object-order vs. image-order rendering
  - ◆ Various techniques (ray casting, splatting, shear-warp, texture mapping, Fourier volume rendering, etc.)
  - ◆ Various types of combinations (compositing, MIP, first-hit, average, etc.)

Eduard Gröller, Helwig Hauser 28

## Slicing

Eduard Gröller, Helwig Hauser 26

## Types of Combinations

- Overview:
  - ◆ MIP ⇒
  - ◆ Compositing ⇒
  - ◆ X-Ray ⇒
  - ◆ First hit ⇒

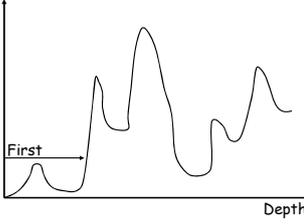
Eduard Gröller, Helwig Hauser 29

# Direct Volume Visualization, Introduction

## Classification – Transfer Functions

Eduard Gröller, Helwig Hauser 30

## First Hit: Iso-Surface Extraction




**First:** Extracts iso-surfaces (again!), done by Tuy&Tuy '84

Eduard Gröller, Helwig Hauser 30

### Average: as X-Ray Images

The graph shows a fluctuating line representing the average intensity across different depths. The X-ray image shows a hand with bones visible, representing the result of an average projection.

**Average:** Produces basically an X-ray picture

Eduard Gröller, Helwig Hauser 31

### Types of Combination

- ◆ Possibilities:
  - ◆  $\alpha$ -compositing
  - ◆ Shaded surface display
  - ◆ Maximum-intensity projection
  - ◆ X-ray simulation
  - ◆ Contour rendering

The diagram shows four small images: NPR (Non-Photorealistic Rendering), x-ray, MIP (Maximum Intensity Projection), and DVR (Direct Volume Rendering). A larger image shows a combination of these techniques, with a semi-transparent colored volume (SSD) overlaid on a DVR image.

Eduard Gröller, Helwig Hauser

### MIP: Maximum-Intensity Projection

The graph shows the maximum intensity value at each depth, resulting in a jagged line. The MIP image shows a branching structure where the most intense pixels are projected, making the structure appear as a solid white mass.

**Max:** Maximum Intensity Projection used for Magnetic Resonance Angiograms, for example

Eduard Gröller, Helwig Hauser 32

### Classification

- ◆ Assignment data  $\Rightarrow$  semantics:
  - ◆ Assignment to objects, e.g., bone, skin, muscle, etc.
  - ◆ Usage of data values, gradient, curvature
  - ◆ Goal: segmentation
  - ◆ Often: semi-automatic resp. manual
  - ◆ Automatic approximation: transfer functions (TF)

**Example**

Eduard Gröller, Helwig Hauser 35

### Compositing: Semi-Transparency

The graph shows an accumulated intensity profile. The resulting image is a semi-transparent volume rendering of a complex structure, allowing internal features to be visible through the outer layers.

**Accumulate:** Make transparent layers visible! Levoy '88

Eduard Gröller, Helwig Hauser 33

### Transfer Functions (TF)

- ◆ Mapping data  $\rightarrow$  "renderable quantities":
  - ◆ 1.) data  $\rightarrow$  color
  - ◆ 2.) data  $\rightarrow$  opacity (non-transparency)

The graph shows a transfer function mapping data values to color and opacity. The x-axis is 'data values' and the y-axis is 'opacity'. A color bar below shows the resulting color gradient from black to red. The function is defined by three segments: 'air' (low data values, low opacity, black), 'skin' (middle data values, semi-transparent, yellow), and 'bone' (high data values, opaque, red).

Eduard Gröller, Helwig Hauser 36

### Different Transfer Functions

- Image results:
  - Strong dependence on transfer functions
  - Non-trivial specification
  - Limited segmentation possibilities

Eduard Gröller, Helwig Hauser 37

### Gradient-Based Transfer Functions

- 2D-Transfer function:
  - Levoy '88
  - Specific opacity at certain threshold
  - but: close-by variation according gradient magnitude
  - highlights transitions (large gradients)
  - dampens homogeneous areas

Eduard Gröller, Helwig Hauser 40

### Lobster – Different Transfer Functions

- Three objects: media, shell, flesh

Eduard Gröller, Helwig Hauser 39

### Multi-Dimensional Transfer Functions (1)

- $f, f', f''$  histograms to depict material boundaries

[Kindlmann, Durkin 1998]  
[Kniss et al. 2002]

Eduard Gröller, Helwig Hauser

### Inclusion of the Gradient

- Emphasis of changes:
  - Special interest often in transitional areas
  - Gradients: measure degree of change (like surface normal)
  - Larger gradient magnitude  $\Rightarrow$  larger opacity

Eduard Gröller, Helwig Hauser 39

### Multi-Dimensional Transfer Functions (2)

- Direct manipulation widgets [Kniss et al. 2002]

1D vs. 2D transfer function

Eduard Gröller, Helwig Hauser 42

## Acknowledgments



- For material for this lecture unit
  - ◆ Roberto Scopigno,  
Claudio Montani (CNR, Pisa)
  - ◆ Hans-Georg Pagendarm (DLR, Göttingen)
  - ◆ Michael Meißner (GRIS, Tübingen)
  - ◆ Torsten Möller
  - ◆ Gordon Kindlmann
  - ◆ Joe Kniss
  - ◆ etc.

