## Spatial Visualization: Grids

CS 5630/6630, Fall 2016 Alexander Lex Guest Lecturer: Aaron Knoll



# The traditional "branches" of visualization (IEEE Visweek)

- Scientific (spatial) Visualization
- Information Visualization
- Visual Analytics



### Scientific visualization

- Data have **spatial** context (usually from simulation or scan)
- Map spatial quantities to colors or geometry, f(space, time) -> rgba
- · 2D or 3D graphics for visualization.



Volume rendering



Tensor field visualization



Map and GIS visualization



Flow visualization





### Information visualization

- Spatial position is secondary or non-existent.
- Illustrate relationships between abstract attributes.



Scatterplots

Charts



## Visual Analytics

- More about interactive user interfaces for data analysis.
- Uses techniques from both scientific visualization and information visualization, as well as statistics, perception, cognition.
- D3+Javascript, R, Matlab
- "Putting it all together"



Management Information Systems (SAS)



Security visualization (Centrifuge)



Genomics (Meyer et al. "Mizbee")







### Scientific Visualization

- "Sci-vis" is about interpreting and rendering spatial data.
- Today:
  - where do spatial data come from? (science domains)
  - how are they represented (*grids*)
  - what can we do with them (*direct* and *indirect* vis)
  - interpolation
- Tues, Nov 17: Volumes Tues, Nov 29: Isosurfaces Thurs, Dec 1: Advanced topics: Vector and Tensor Fields



### Scientific Visualization

- Data sources
- Data representation
  - fields
  - grids
- Data interpretation
  - The scientific visualization pipeline
  - Interpolation



### Data sources



## Computational Data

#### The output of scientific computing:

physics, chemistry, blood flow, neurophysiology, meteorology, climatology, astronomy...



- Nuclear physics
- Quantum chemistry
- Molecular dynamics
- Computational fluid dynamics
- Rigid-body and structural mechanics
- Coarse-grained dynamics, agents simulations
- Meteorology
- Astrophysics
- Cosmology



### Scanned data

### **The output of instruments** in medical imaging, microscopy, telescopy, GIS

- X-ray crystallography
- Synchrotron / radiation light sources
- Transmission electron microscopy
- Confocal microscopy
- Camera imagery
- Ultrasound
- Magnetic resonance imaging
- X-ray tomography
- Satellite
- Telescope

#### Angstroms















# Data representation (grids)



### Fields

• Mathematically, a *field* is a set of elements with addition, multiplication operators that satisfy the field axioms

name	addition	multiplication
associativity	(a+b)+c = a + (b+c)	(a b) c = a (b c)
commutativity	a+b=b+a	a b = b a
distributivity	$a\left(b+c\right) = ab+ac$	(a+b) c = a c + b c
identity	a + 0 = a = 0 + a	$a \cdot 1 = a = 1 \cdot a$
inverses	a + (-a) = 0 = (-a) + a	$a a^{-1} = 1 = a^{-1} a$ if $a \neq 0$
		wolfram.com

• Intuitively, a field is a varying quantity defined continuously over space.



### Fields

#### with a 2D domain





#### vector field



#### tensor field





### Multifields



Radiation hydrodynamics in Enzo: Joe Insley (ANL), Rick Wagner (SDSC) <u>https://vimeo.com/17771397</u>

### Multifields



8 molecular orbitals of a LiAlH2O DFT simulation, courtesy Aiichiro Nakano, University of Southern California

- dimension of domain (the field)
- dimension of the data to visualize (the geometry)



- dimension of domain (the field)
- dimension of the data to visualize (the geometry)



- dimension of domain (the field)
- dimension of the data to visualize (the geometry)



- dimension of domain (the field)
- dimension of the data to visualize (the geometry)



- dimension of domain (the field)
- dimension of the data to visualize (the geometry)



- dimension of domain (the field)
- dimension of the data to visualize (the geometry)



- dimension of domain (the field)
- dimension of the data to visualize (the geometry)



- dimension of domain (the field)
- dimension of the data to visualize (the geometry)



- dimension of domain (the field)
- dimension of the data to visualize (the geometry)



- dimension of domain (the field)
- dimension of the data to visualize (the geometry)



- dimension of domain (the field)
- dimension of the data to visualize (the geometry)



### Grids

- Continuous fields are an illusion
- All data are discrete
- Meshes are chosen based on what is computationally efficient for the
- Visualization software must implement data models to handle a wide range of field and non-field data



### Structured vs Unstructured

- In general, from the relational database world:
  - Structured data are data that are indexed, and can be accessed via a hash, array, or other query.
    - I.e., search time *O*(1) or *O*(*log N*).
  - **Unstructured data** are not indexed you have to brute-force search to find them.
    - I.e. search time O(N)
- In information/data visualization:
  - structured means data you've already indexed, organized (for example, in D3).
  - **unstructured** is everything else (i.e. text, imagery, video, foo) you have to search through.
- In scientific visualization, this can get a bit confusing...
- First we need to differentiate between geometry and topology.



# Geometry vs Topology

- Geometry
  - Position of vertices in Euclidean space
  - Can be uniform, structured or unstructured.
- Topology
  - Defines the "cells", or connectivity of the vertices.
  - Can also be structured or unstructured.



# Uniform grid geometry

- Uniform spacing along the axes, also known as "raster data".
- Most volume data look like this; structured data usually means this.
- You still need metadata to know the size of the axes!







- Representation of scalar 3D data set  $\Omega \in R^3 \to R$
- Analogy: pixel (picture element)



- Voxel (volume element), with two interpretations:
  - Values between grid points are resampled by interpolation



- Collection of voxels
- Uniform grid



### Arecibo Message

- Way of understanding mechanics of raster image representation
- Radio telecope in Puerto Rico
- built in 1964, renovated in 1974
- To celebrate: Frank Drake and Carl Sagan (Cornell University) sent message to M13 in Hercules (25,000 light years away)
- 1679 bits, frequency modulate 2380 MHz



### The Message

#### 1679 bits were encoded as 2380MHz plus and minus some frequency

 $1\,1\,1\,0\,0\,0\,0\,1\,1\,1\,0\,0\,0\,0\,0\,1\,1\,0\,1\,1\,0\,0\,0\,0\,0\,0\,0\,0\,0\,0\,1\,1\,1\,0\,0\,1\,0\,0\,0\,0\,0\,0\,0\,0\,0\,0\,0\,0\,1\,1\,1\,1\,1\,1\,0\,0\,1\,0\,0$ 01111001111101001111000

This is a **1-D** sequence of bits in time How will an alien understand this list of bits? (will have different symbols than "0" "1") No meta-information!

### Understanding the message

- Perhaps some "visual" representation of bits
  - (what is black vs white?)
- Aliens notice 1679 = 23 x 73 (product of two primes)
- Perhaps its not a linear sequence: 2-D array
- •Two ways of sequencing values in 2D array
- •Various ways of laying them out in 2D space
- •Then: have to decipher it!

# 73 x 23 compare to:



compare to: http://en.wikipedia.org/wiki/Arecibo\_message

23 x 73: what was different?


#### Structured (rectilinear) grid geometry

- Still structured, but with non-uniform spacing along the axes.
- Positions can still be computed procedurally
- Some meteorology, climate CFD data like this.



$$P_{i,j,k} = P_{0,0} + \Delta_x[i]\vec{e}_x + \Delta_y[j]\vec{e}_y$$







## Unstructured geometry

- Raw, unstructured point data.
- You actually need to store the x,y,z positions of vertices.
- Some of the largest computational and scanned data
  - LiDAR, RGB-D point clouds, range scans
  - n-body codes molecular dynamics, cosmology
- Note: this is unstructured topology, too!





30-billion particle "Cosmic Web" data — Paul Shapiro, University of Texas at Austin I Wald, A Knoll, G.P. Johnson, W. Usher, V. Pasccci & M.E. Papka: "CPU Ray Tracing Large Particle Data with P-k-d Trees" IEEE Visualization 2015



## Structured grid topology

- You can have unstructured geometry but structured topology
  - Implicit definition of cells
  - Implicit connectivity between vertices
- More exotic options with structured grid topology:
  - Finite elements, finite differences on curvilinear grids
  - spectral F/E, some spline-based finite elements simulations
  - Good for precision-critical flow computations (blood flow, CFD)







## Curvilinear grids



Image:T.U. Graz



#### Spectral/hp finite elements



Data: George Karniadakis, Brown University. Visualization: Joe Insley, ANL



#### Spectral/hp finite elements



Data: George Karniadakis, Brown University. Visualization: Joe Insley, ANL



### Unstructured grid topology

- Both uniform elements and "mixed elements" (allowing any cell type)
- Need to store vertices and indices separately
  - mixed elements: vertices, indices and count
- Many, many finite elements codes.
  - solid mechanics, CAD
  - bioelectric modeling







#### Finite elements



Visualization: Steve Owen, using Cubit













## Geometry vs topology

		Topology	
		Structured	Unstructured
Geometry	Uniform	Image	Unstructured
	Structured	Rectilinear	Unstructured
	Unstructured	Curvilinear	Unstructured



## Colloquially

- In spatial (scientific) vis we usually talk about geometry, and:
  - **structured** means rectilinear grid (usually, but not always uniform).
  - **unstructured** means everything else (curvilinear grids, tetrahedra, hexahedra, points, etc.)



















## Non-field and other data

- In addition to structured/unstructured field data, you can have non-field geometry.
  - Boundary surface meshes
  - Atom positions, bonds, ribbons
- Non-geometric annotations
  - Especially in GIS.
- Visualization data models are complex!









#### What do we do with these data?

- In computer graphics, life is "easy"
  - Have a triangle mesh, render it!
- Visualization is more than just rendering.
- Two approaches:
  - direct visualization:
    i.e. render from a (usually 3D) field directly
  - indirect visualization:
    i.e. convert the field to triangles and render those (usually with GPU rasterization)



### Indirect visualization





### Direct visualization





#### The indirect-direct spectrum

- Most data must be processed somewhere after acquisition
- Direct/Indirect are not "absolute", but relative.
  - Direct = less processing prior to rendering
  - Indirect = more processing prior to rendering
- Multiple questions:
  - how do the data need to be transformed for the desired analysis?
  - what does the target rendering engine / API require?
- There is no such thing as pure "direct" or "indirect" visualization; only "more direct" or "more indirect".



#### The indirect-direct spectrum

isosurface extraction (marching cubes) + rasterization

direct isosurface ray casting

volume rendering

from raw data

segmentation+filtering+ classification+rasterization pipeline

polygonal

splatting

Indirect





Direct



## The visualization pipeline

• Even if we merge filtering and rendering, it is still helpful to think of them as a chain of operations.



• Visualization workflows take the form of a flow chart, tree or network...



#### SCIRun





#### ParaView





#### VisTrails





### VTK

- <u>http://www.vtk.org</u>, open-source, developed and maintained by Kitware.
- The standard-bearer API for general-purpose scientific visualization
  - Full-fledged data model for structured, unstructured, particle data
  - Marching cubes, cut/clip planes, streamlines, etc.
  - Hundreds of other analysis filters
  - Numerous readers for common scientific formats
- Call as a library from C++, Java, Python, Tcl/Tk
- Limitations:
  - no UI you need to code (or at least, script) your workflows.
  - data model can be "heavy", memory-inefficient but it nearly always works!





S. Bruckner, "Data Structures in the Visualization Toolkit."



## Attribute types in VTK



(5) Tensor S. Bruckner, "Data Structures in the Visualization Toolkit."



## Simple data flow in VTK



Figure 7: Resulting image for program 3.2



Figure 8: Data-flow chart for program 3.2



S. Bruckner, "Data Structures in the Visualization Toolkit."

## Interpolation



## Interpolation

- Converting from discrete to continuous i.e. grid to field.
- How do we find the values of points "inside" a grid/mesh?
  - Indirect visualization: how to find vertices of triangles
  - Direct visualization: how to find the value of samples in space
- i.e., from an explicit grid, we want to evaluate a field **f** at some point **x**
  - The way we do this interpolation is called the filter kernel of the field f.



## Mesh Choice Impacts How the Continuous Data is Interpreted

- Two key questions:
  - Sampling, or the choice of where attributes are measured
  - Interpolation, or how to model the attributes in the rest of space



#### Interpolation

- **Continuous** reconstruction of **discrete** input data  $F:\mathbb{R}^n \to \mathbb{R}^n$   $(\mathbf{x}_i, f_i) \to \mathsf{value}$  $\forall i \in \{1, ..., n\}, F(\mathbf{x}_i) = f_i$
- Depends on grid structure (when available)
- Interpolation vs. approximation

- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?



- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?



- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?



- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?
  - I[1.3] = I[round(1.3)] = I[1]


#### Nearest Neighbor Interpolation

- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?
  - I[1.3] = I[round(1.3)] = I[1]



- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?



- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?



- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?



- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?



- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?
  - Let s = 1.3 round(1.3)



- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?
  - Let s = 1.3 round(1.3)
  - I[1.3] = 0.7\*I[1] + 0.3\*I[2] = (1-s)\*I[1] + s\*I[2]



- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?
  - Let s = 1.3 round(1.3)
  - I[1.3] = 0.7\*I[1] + 0.3\*I[2] = (1-s)\*I[1] + s\*I[2]



#define **lerp(a,b,t)** (1-t) \* a + t\*b

#### Bilinear interpolation

Just 3 linear interpolations

#define lerp(a,b,t) (1-t) \* a + t\*b

//Given voxel vertices cXX and the (tx,ty) position within the voxel [0,1]^2

//lerp along y direction
float  $c00_y = lerp(c00, c01, ty);
float c10_y = lerp(c10, c11, ty);$ 

//lerp along x direction
return lerp(c00\_y, c10\_y, tx);



© www.scratchapixel.com

#### Bilinear interpolation





#### Trilinear interpolation

Just 7 linear interpolations!

#define lerp(a,b,t) (1-t) \* a + t\*b

//Given voxel vertices cXXX and the (tx,ty,tz) position within the voxel [0,1]^3

//lerp along z direction. float  $c000_z = lerp(c000, c001, tz)$ ; float  $c010_z = lerp(c010, c011, tz)$ ; float  $c100_z = lerp(c100, c101, tz)$ ; float  $c110_z = lerp(c110, c111, tz)$ ;

//lerp along y direction float  $c000_yz = lerp(c000_z, c010_z, ty);$ float  $c100_yz = lerp(c100_z, c110_z, ty);$ 

//lerp along x direction
return lerp(c000\_yz, c100\_yz, tx);



© www.scratchapixel.com

## Isosurfaces from trilinear filter kernels





Another formula for trilinear interpolation (3D)

$$f(x, y, z) = \sum_{i, j, k \in \{0, 1\}} x_i y_j z_k v_{ijk}$$

Where  $x_0 = i + 1 - x$ ,  $x_1 = x - i$ , ditto for y and z And  $v_{ijk}$  is the value of the voxel at that vertex.

#### General interpolation (3D)

$$f(x, y, z) = \sum_{i,j,k} B_i(x) B_j(y) B_k(z) v_{ijk}$$

Where the B(x) is a general basis function, v is the voxel.

# Trilinear vs B-spline filtering for volume rendering

#### trilinear tri - cubic-B-spline



### Next Tuesday lecture

• Volume rendering.

