A close-up photograph of a person's hand interacting with a 3D brain scan visualization on a computer monitor. The hand is pointing at a specific area of the brain model, which is color-coded in various shades of blue, green, and yellow. The monitor displays multiple slices of the brain scan in a grid. In the top right corner of the slide, there is a large, semi-transparent title.

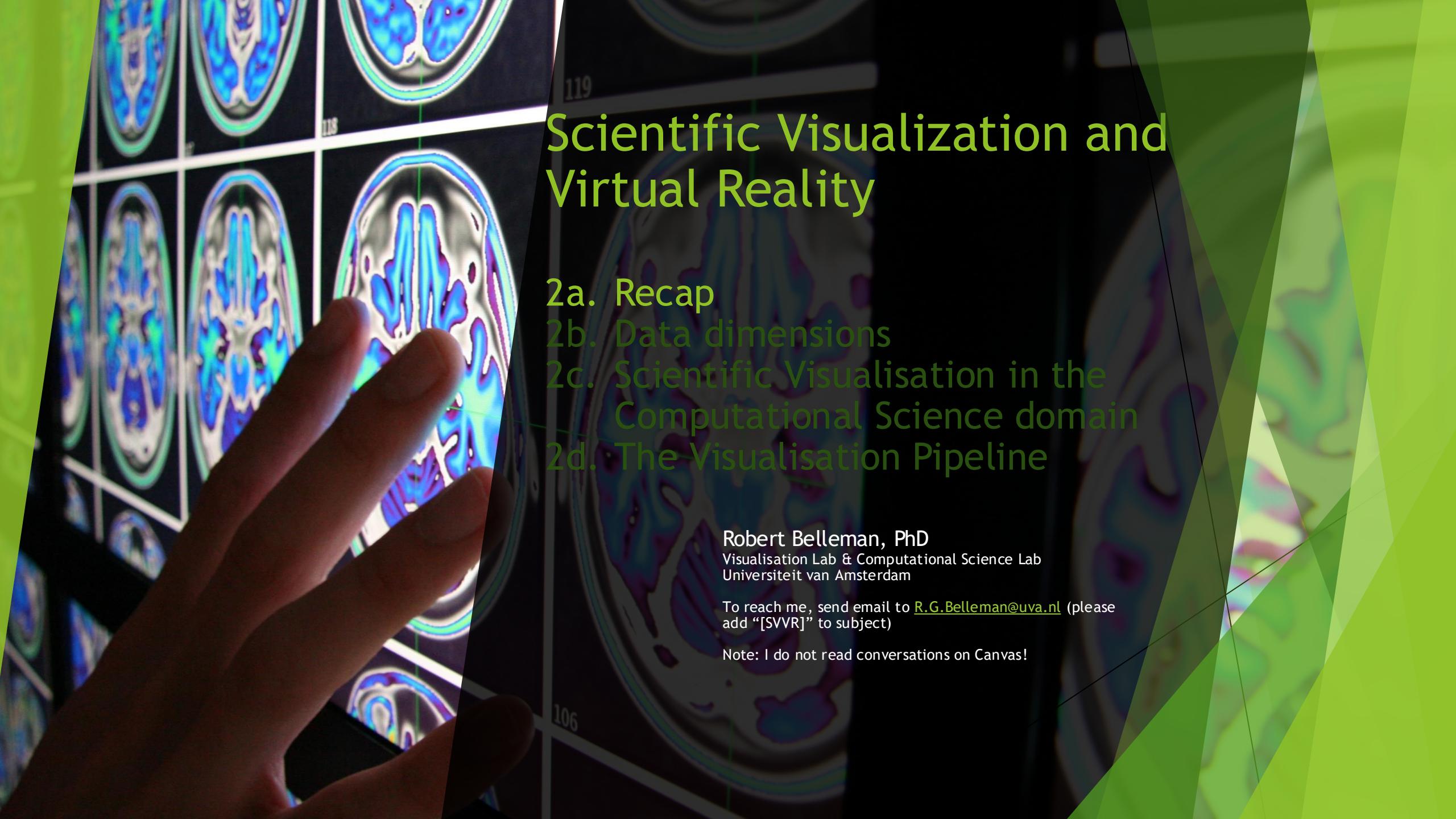
Scientific Visualization and Virtual Reality

- 2a. Recap
- 2b. Data dimensions
- 2c. Scientific Visualisation in the Computational Science domain
- 2d. The Visualisation Pipeline

Robert Belleman, PhD
Visualisation Lab & Computational Science Lab
Universiteit van Amsterdam

To reach me, send email to R.G.Belleman@uva.nl (please add “[SVVR]” to subject)

Note: I do not read conversations on Canvas!

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Bertin's “Levels of organization”

Position

N	O	Q
N	O	Q
N	O	Q
N	o	
N		
N		
N		
N		

Size

Value

Texture

Colour

Orientation

Shape

N Nominal

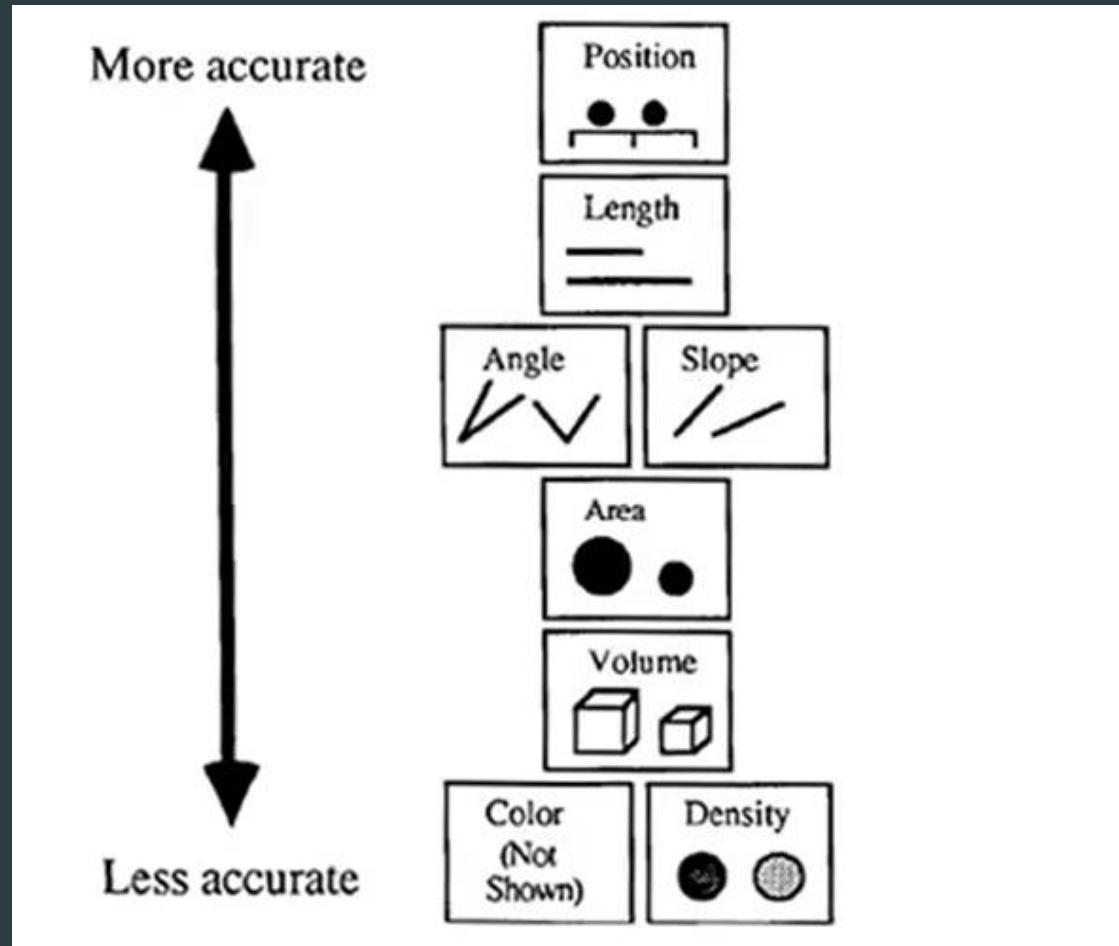
O Ordered

Q Quantitative

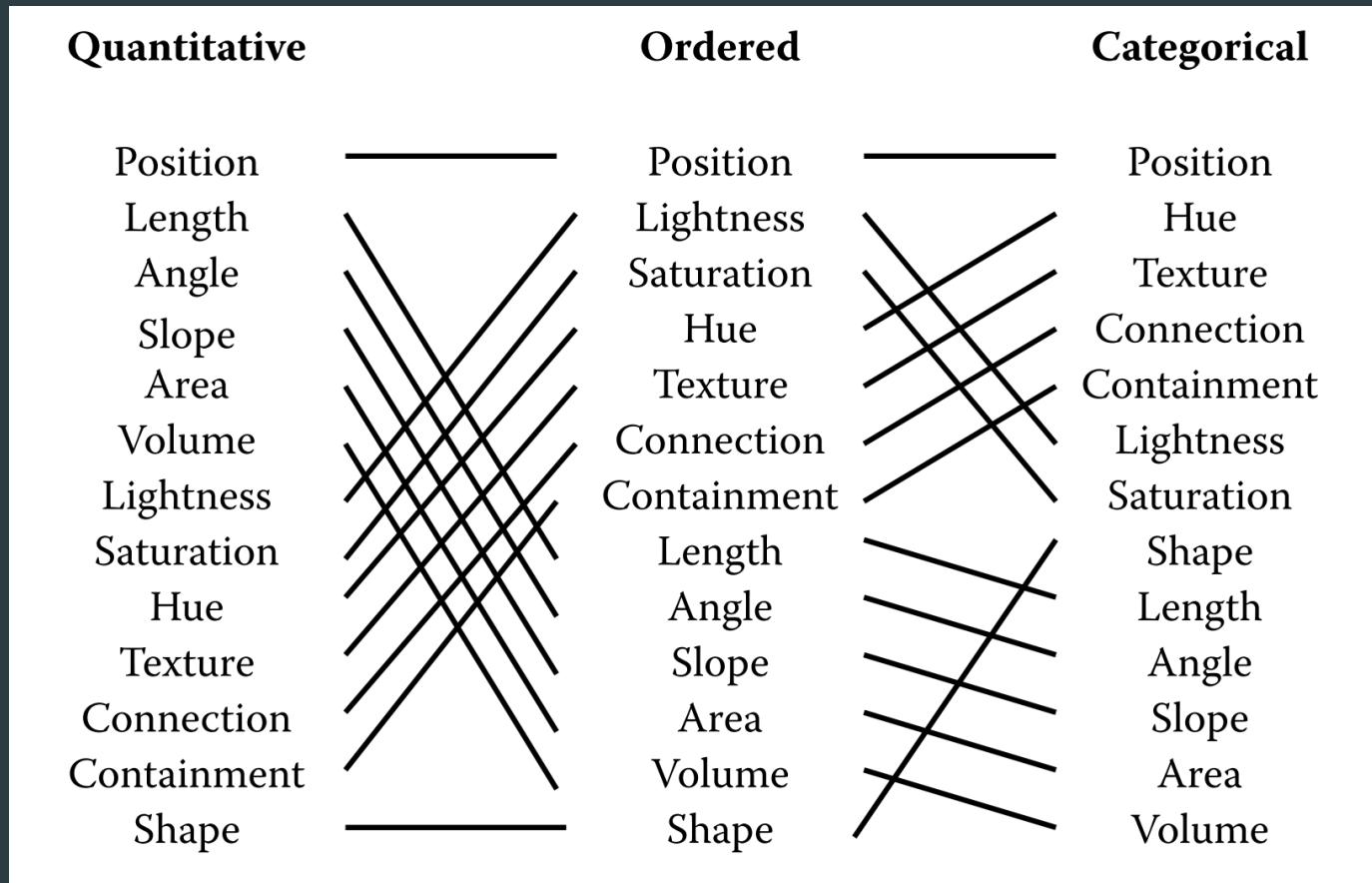
Note that Bertin disregarded:

- Position in 3D
- Embedding/containment
- Time (animation)
- Interaction

Mackinlay's “Perceptual properties”



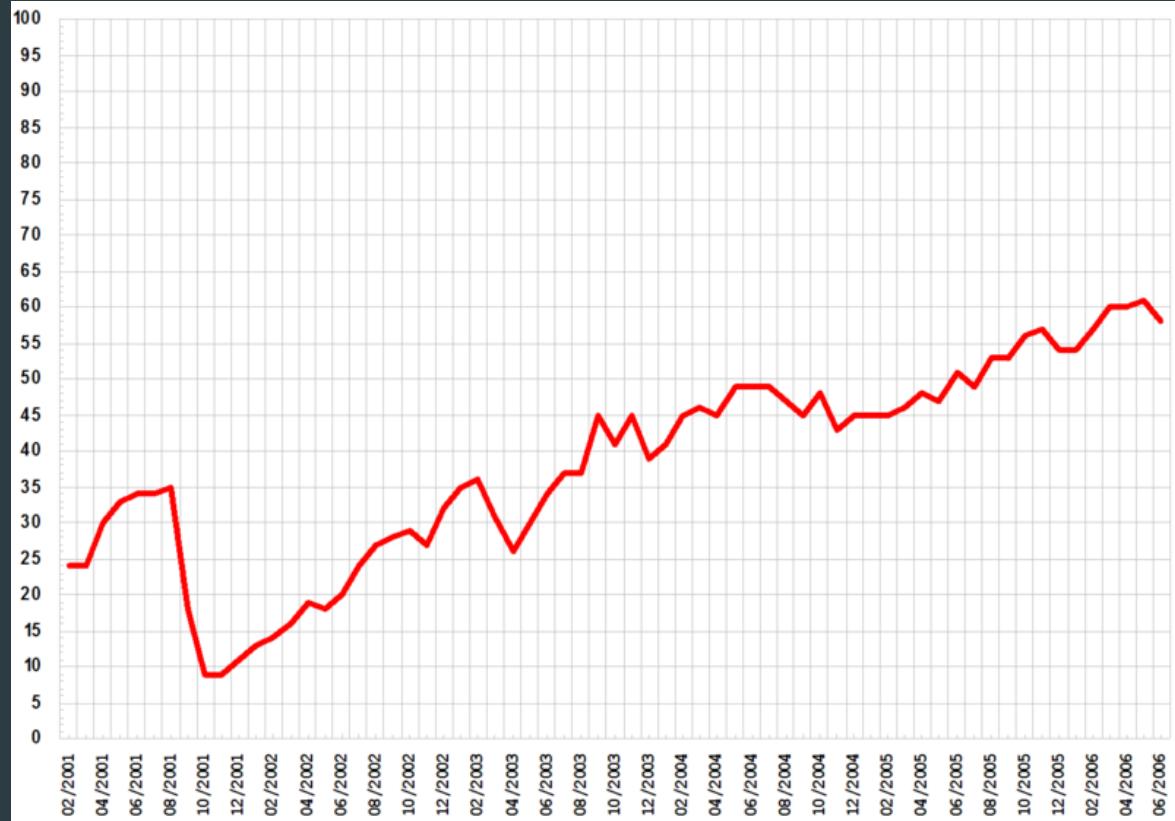
Perceptual properties



Our ability to perceive information encoded by a visual channel depends on the type of data used, from most accurate at the top to least at the bottom. Redrawn and adapted from (Mackinlay, 1986)

Source: Jock Mackinlay. 1986. Automating the design of graphical presentations of relational information. ACM Trans. Graph. 5, 2 (April 1986), 110-141.
<https://doi.org/10.1145/22949.22950>

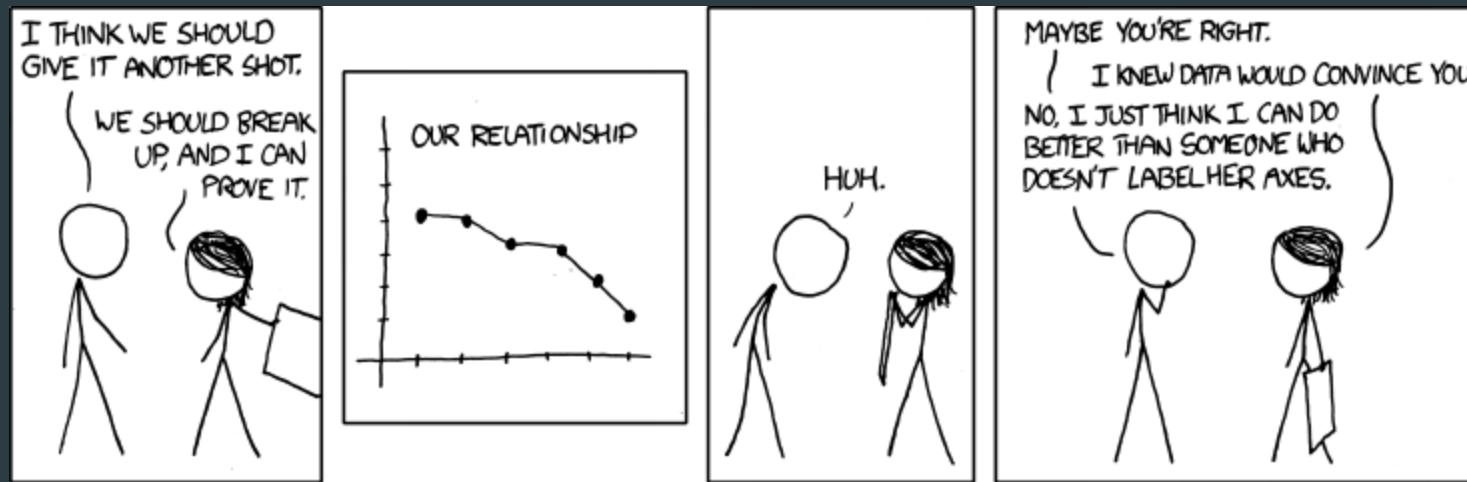
“Bush disapproval ratings”



- ▶ X-position: time (Q)
- ▶ Y-position: percentage (Q)

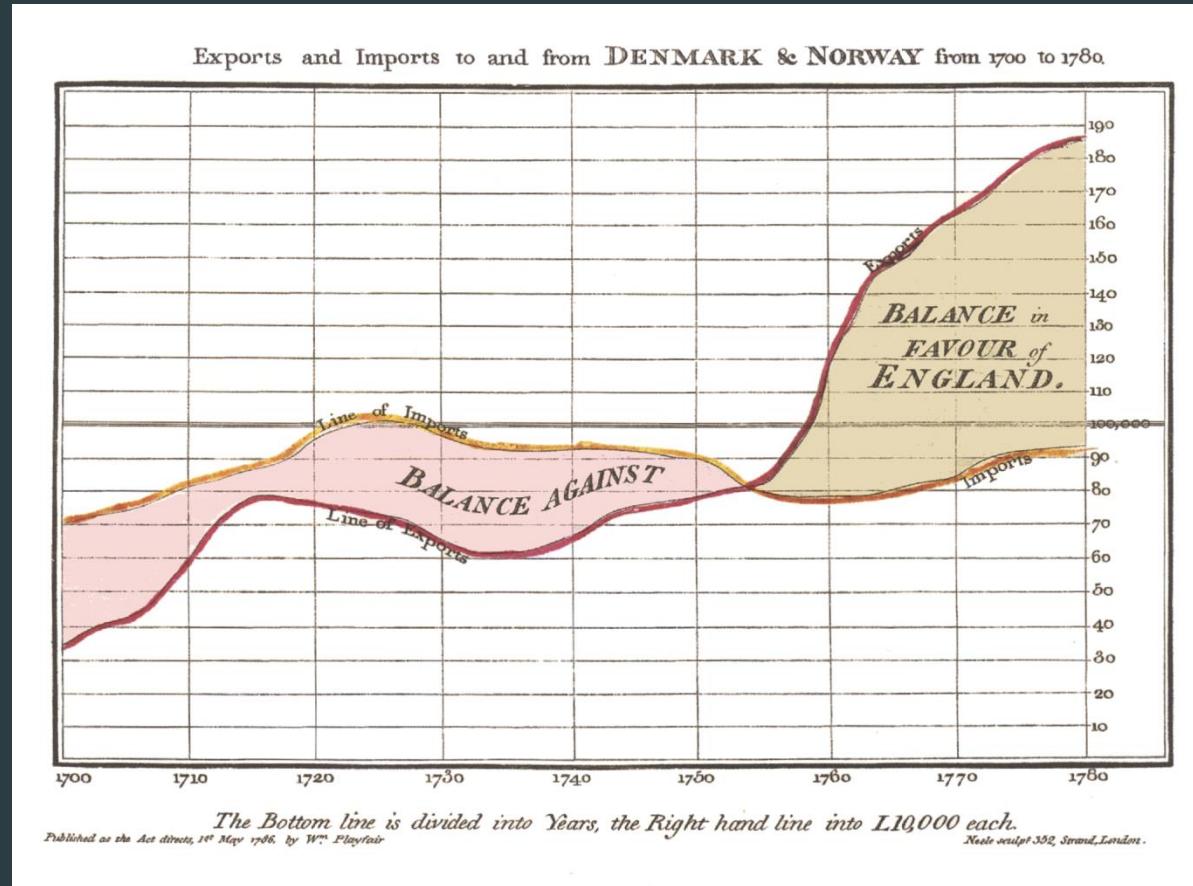
Don't forget!

- ▶ Label axes!
- ▶ Add keys, ticks, title!



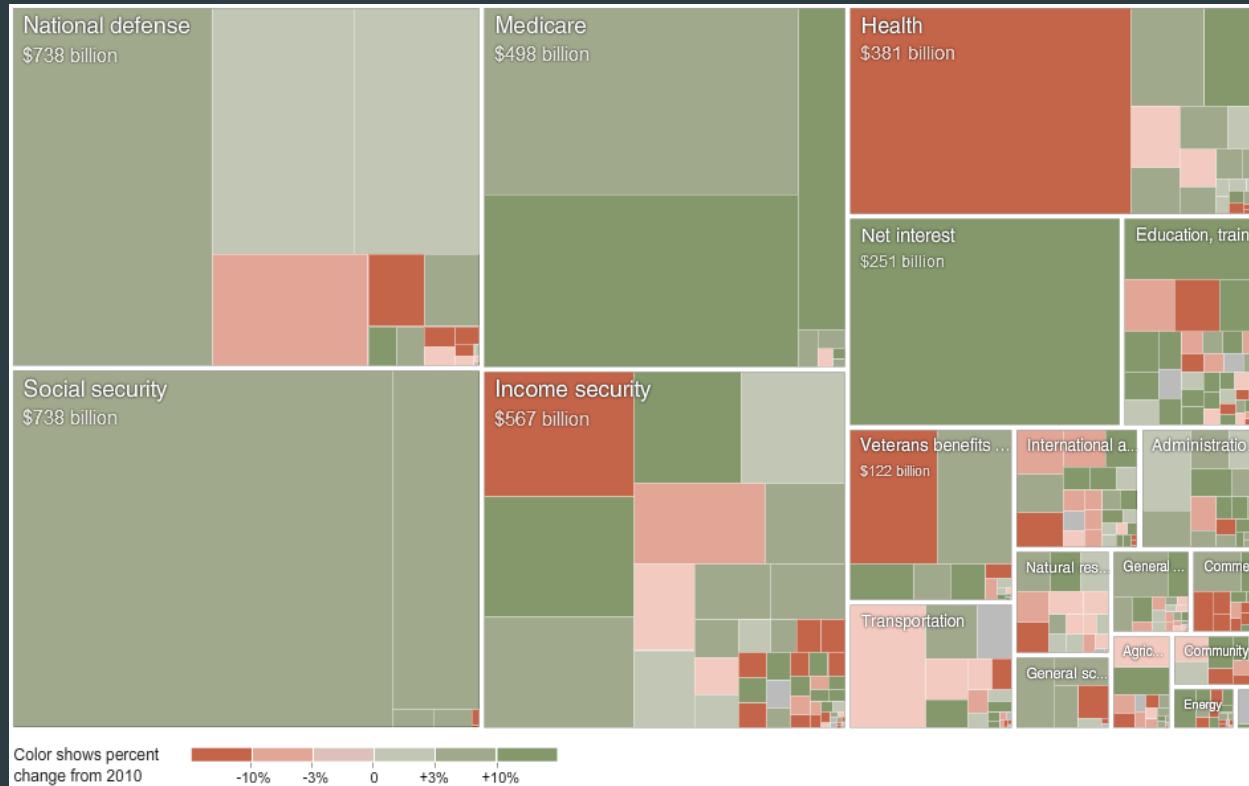
Source: <http://xkcd.com/833/>

William Playfair (1786)



- ▶ X-position:
year (Q)
- ▶ Y-position:
imports/exports
currency (Q)
- ▶ Colour and area:
balance of trade (N, O)

Obama's 2011 budget proposal



- ▶ Rectangle position: category (N)
- ▶ Rectangle area: amount (Q)
- ▶ Color hue: direction of change (N, O)
- ▶ Color value: magnitude of change (Q)

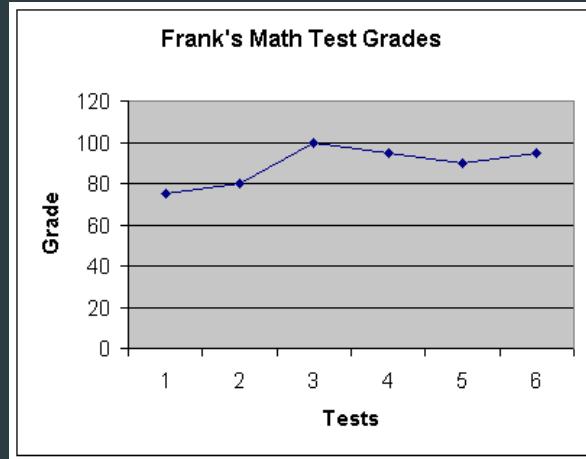
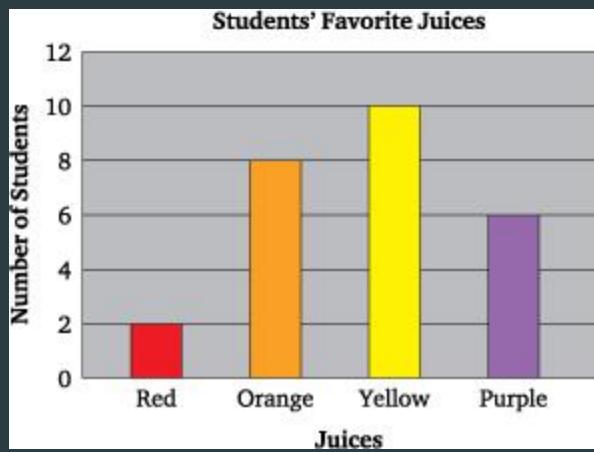
A close-up photograph of a person's hand interacting with a 3D brain scan visualization on a computer monitor. The hand is pointing at a specific area of the brain model, which is color-coded in various shades of blue, green, and yellow. The monitor displays multiple slices of the brain scan, with numerical labels (119, 106) indicating the slice number. The background is dark, making the colorful brain scan stand out.

Scientific Visualization and Virtual Reality

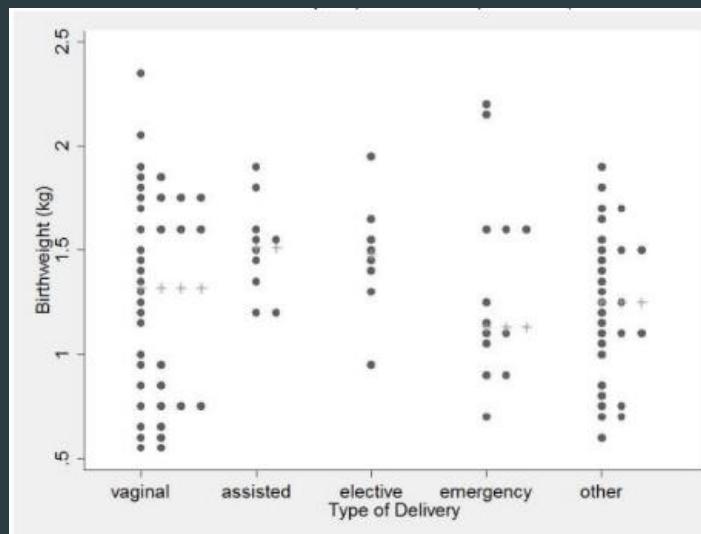
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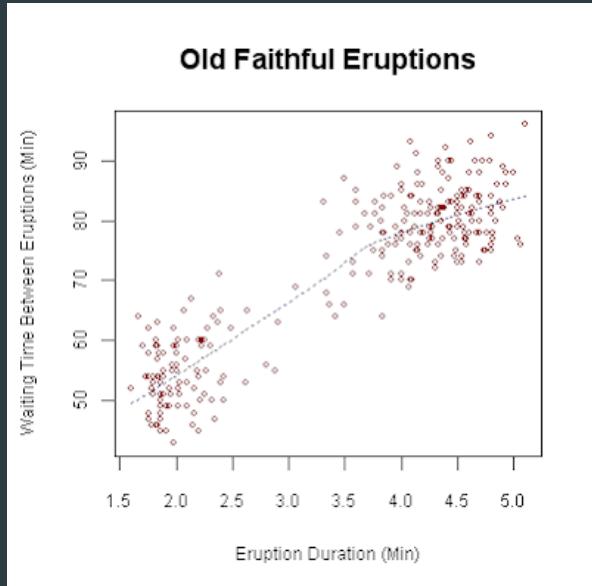
Univariate data



	A
1	
2	
3	
...	



Bivariate data

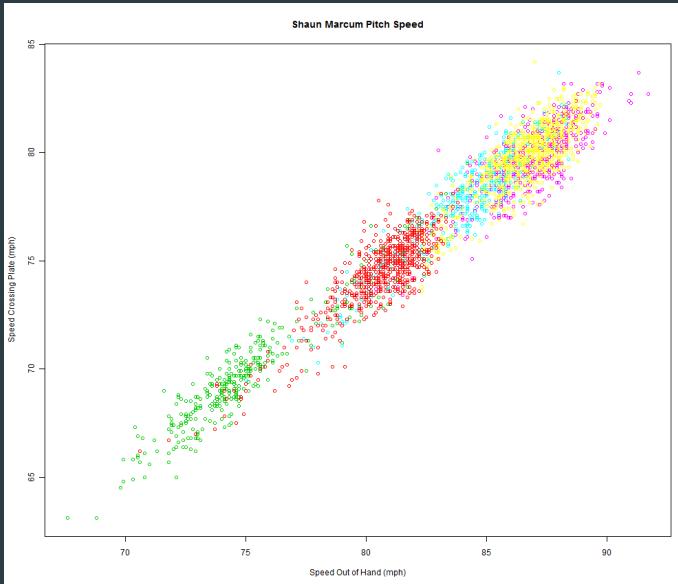


2D scatter plot ([source](#))



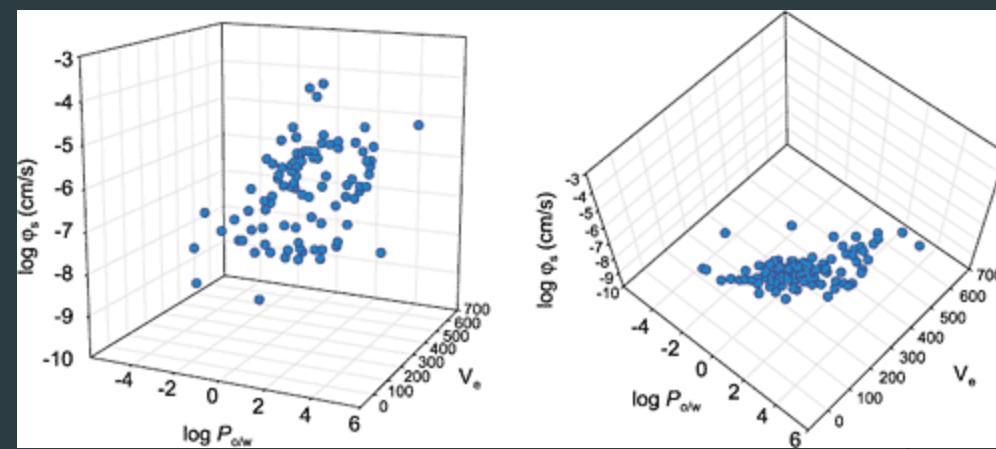
	A	B
1		
2		
3		
...		

Trivariate data



2D scatter plot with
third variable mapped
to colour

	A	B	C
1			
2			
3			
...			



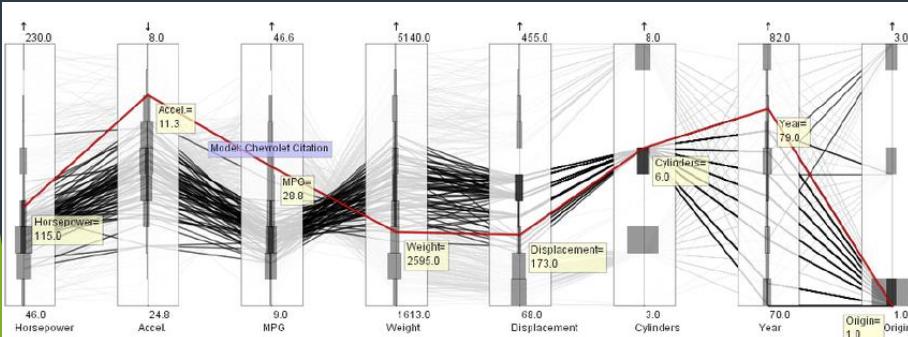
3D scatter plot?

Multidimensional data

How many variables can be depicted in an image?

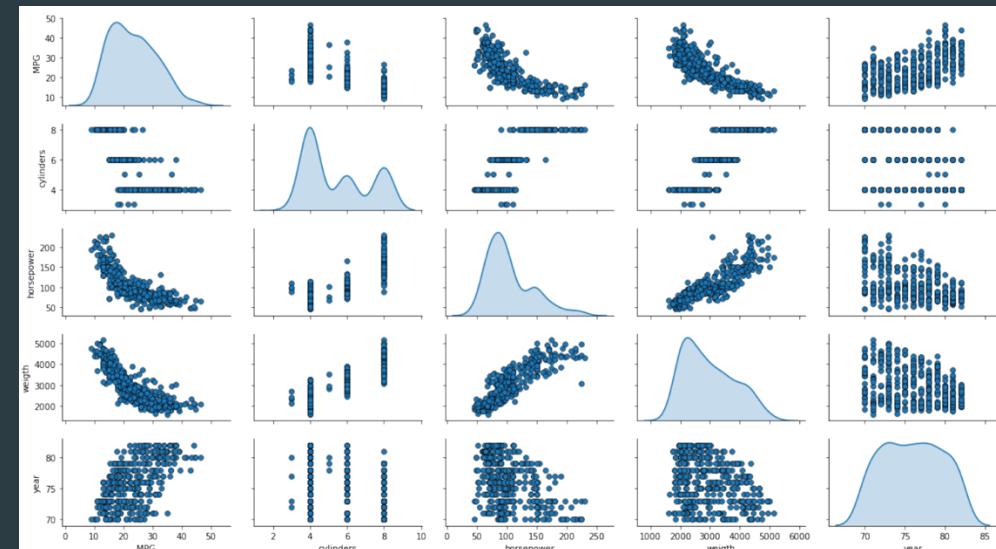
"With up to three rows, a data table can be constructed directly as a single image... However, an image has only three dimensions. And this barrier is impassible."

Bertin



Parallel coordinates

	A	B	C	...
1				
2				
3				
...				



Pair plots / scatterplot matrix / [HyperSlice](#)

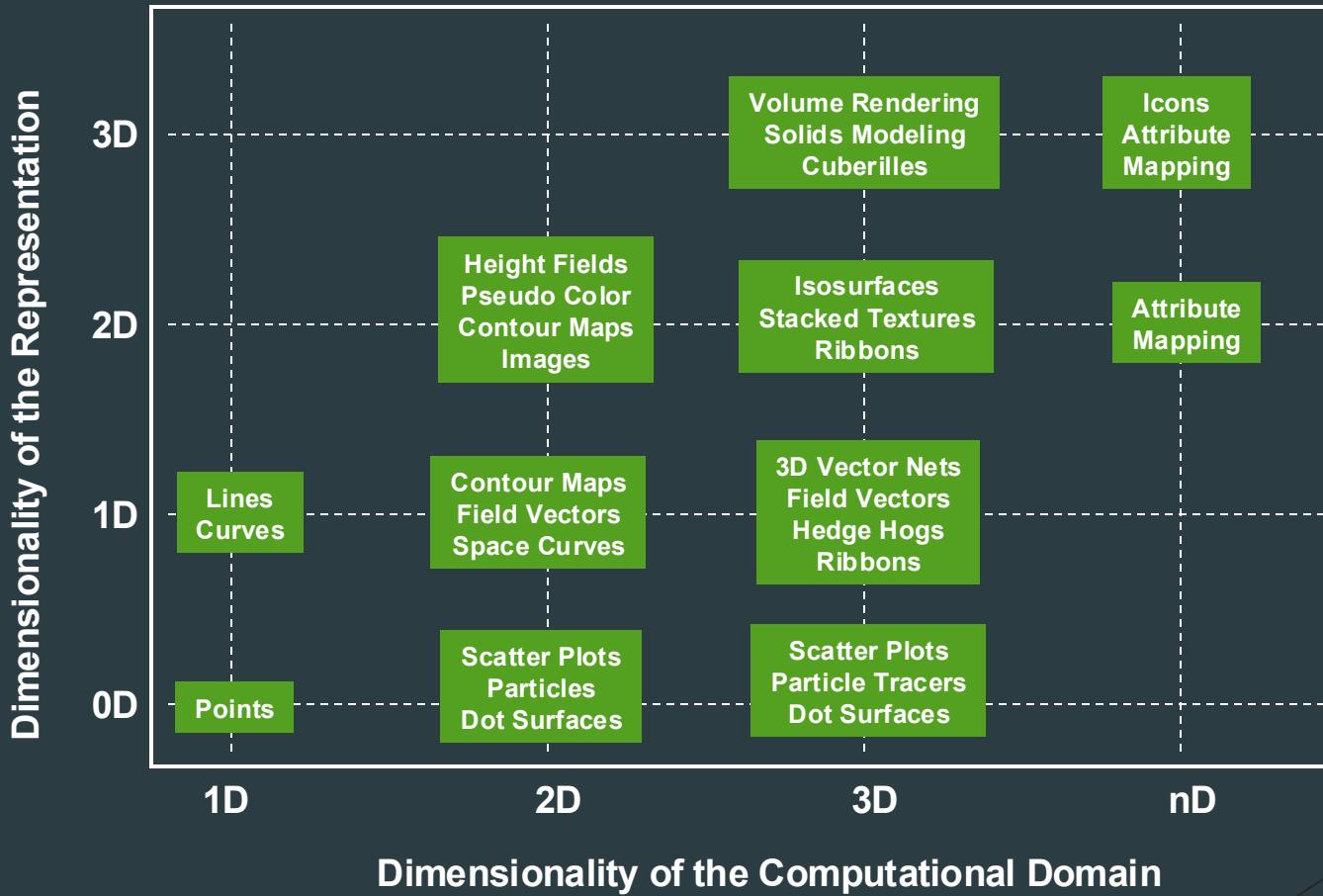


COVER

- move world
- scale world
- view all
- reset view
- stop headtracking
- navigation...
- view options...
- ClipPlane...
- File Browser
- ModelLoader
- PanoView360
- Point Model Plugin
- Volume
- ImagePlugin
- CAMERAVis
- Floorplan Viewer
- PDB Viewer
- Molecules...

- Protein size
- Layout Menu
- Show name
- Show Topsan Info
- Sequence Viewer
- Clear Alignment
- Delete All

Visualization mapping space



Based on: Craig Upson, *Volumetric Visualization Techniques*, in "State of the Art in Computer Graphics: Visualization and Modeling", David F. Rogers Rae A. Earnshaw (eds.), Springer-Verlag, 1991, pp. 313-350

Effective visual encoding

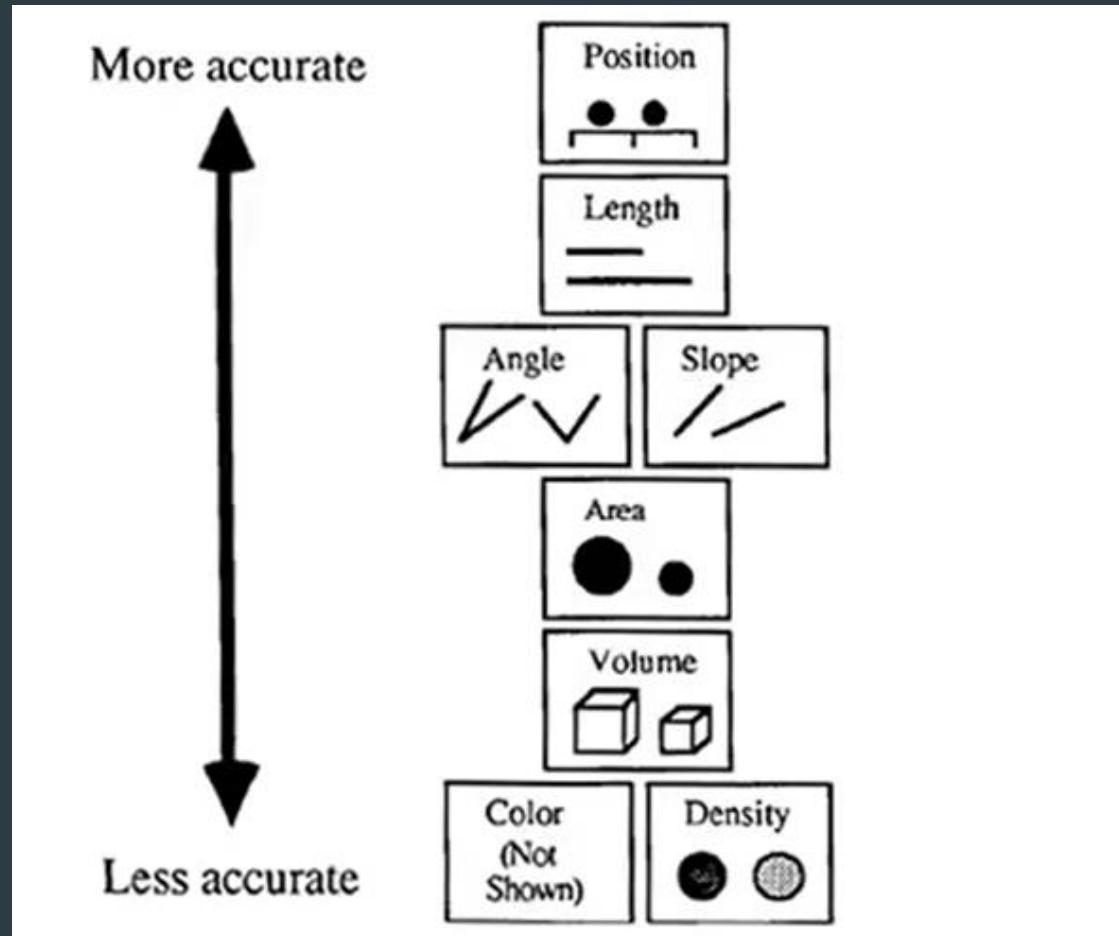
Challenge:

- ▶ Choose the best encoding (or mapping) from data to visual attributes

Consider:

1. **Importance ordering:** encode the most important information in the visually most accurate way
2. **Expressiveness:** show all data and *only* the data
3. **Consistency:** the properties of the visual attributes must match the properties of the data

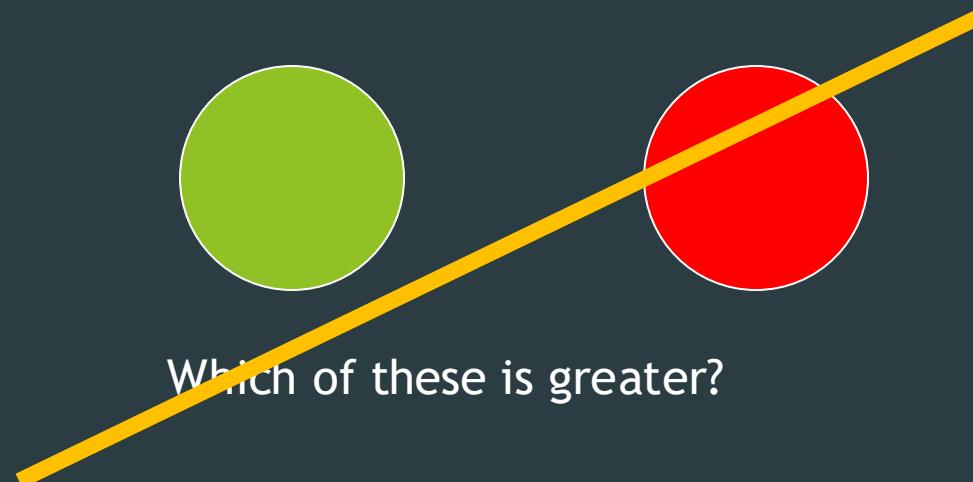
Importance ordering: perceptual properties



Expressiveness

Mackinlay:

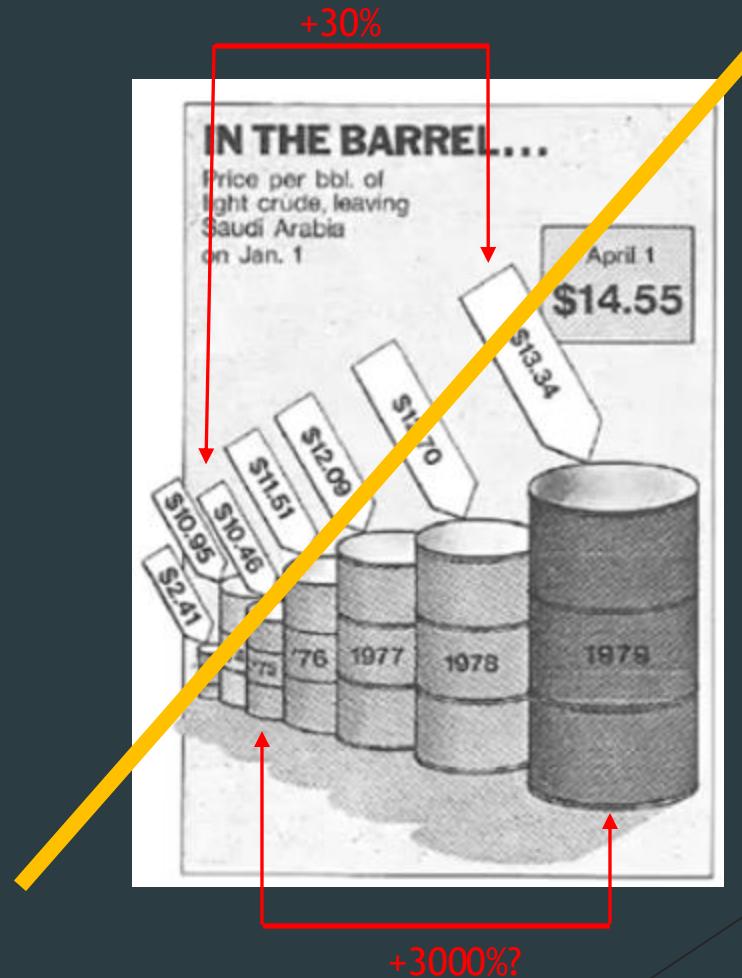
- ▶ “A set of facts is expressible in a visual language if the sentences (i.e. the visualizations) in the language express all the facts in the set of data, and *only* the facts in the data.”



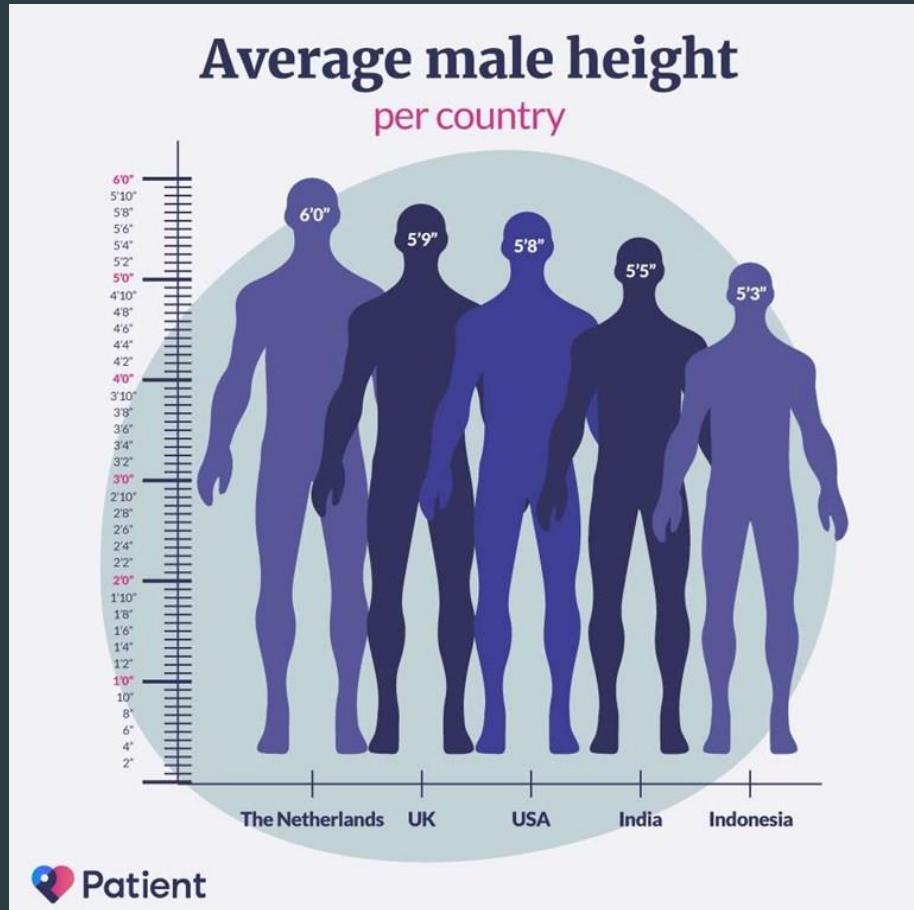
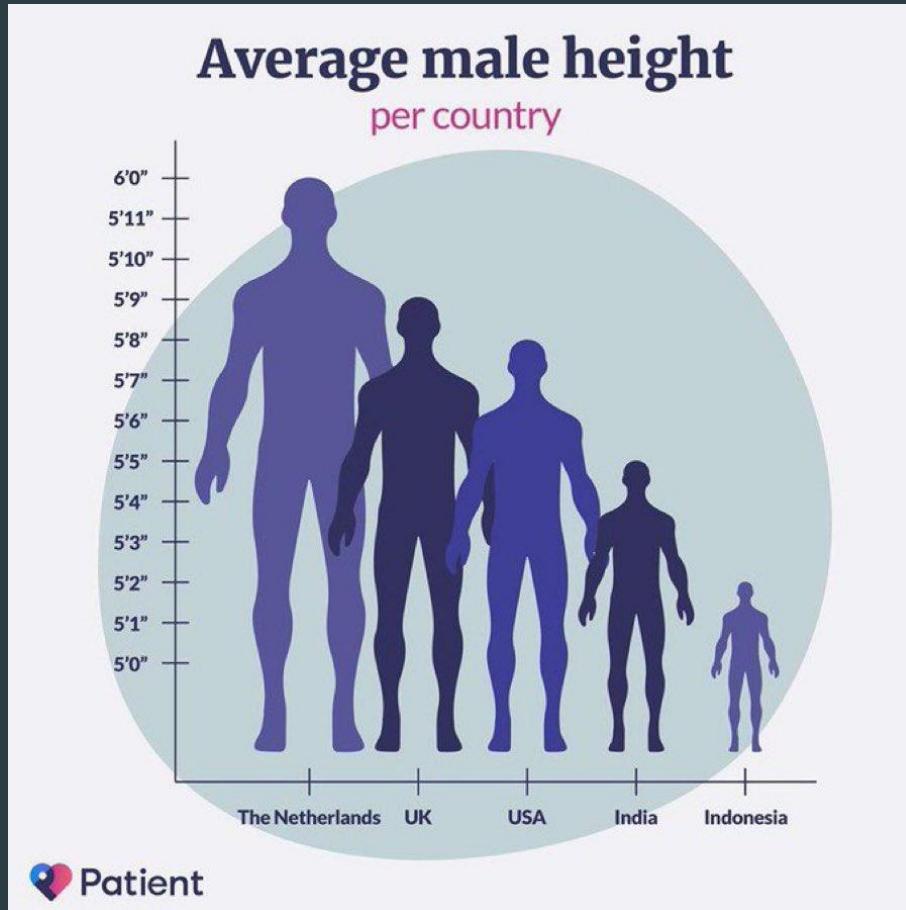
Visual consistency

- ▶ The properties of the visual attributes must match the properties of the data

E.g. don't map one-dimensional data to two- or three-dimensional representations!



Visual consistency

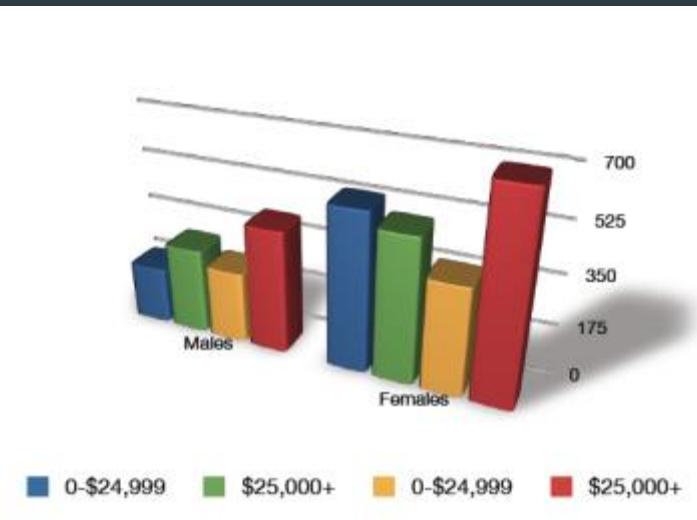


Data-to-ink ratio

- Maximize Data-to-ink ratio =
$$\frac{\text{Data Ink}}{\text{Ink used in graphic}}$$



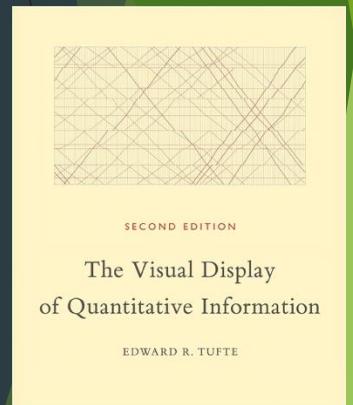
High data-to-ink ratio



Low data-to-ink ratio



Edward Tufte



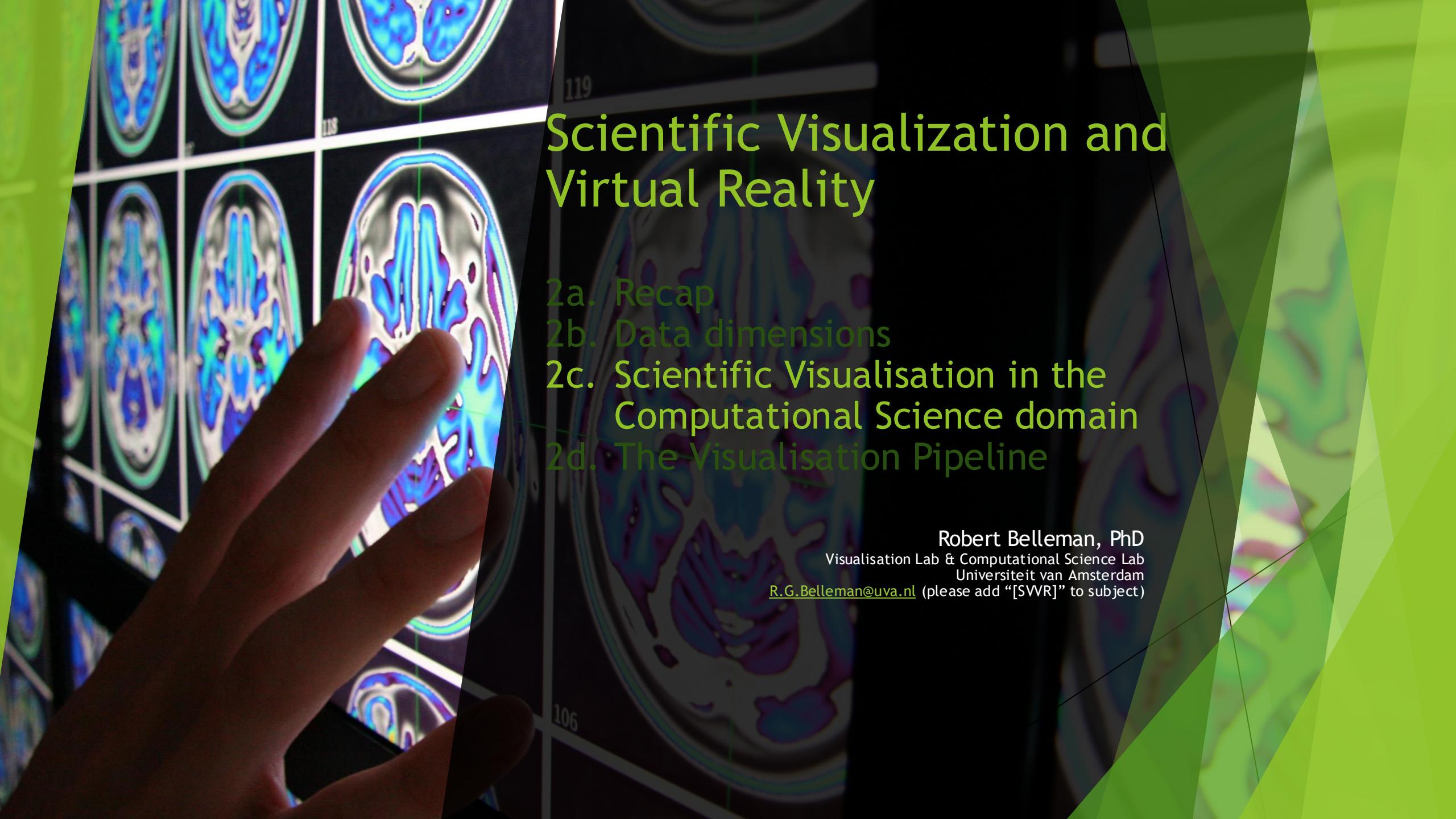
Esthethics



Photo: [Michael Najjar](#)

Remember this

1. All representations of information are subjective interpretations of the information, not the information itself.
2. The challenge in visualization is to find a representation that is both efficient and appropriate.
3. The purpose of visualization is to inform, not to misinform.
4. Any visualization is limited in its interpretation.

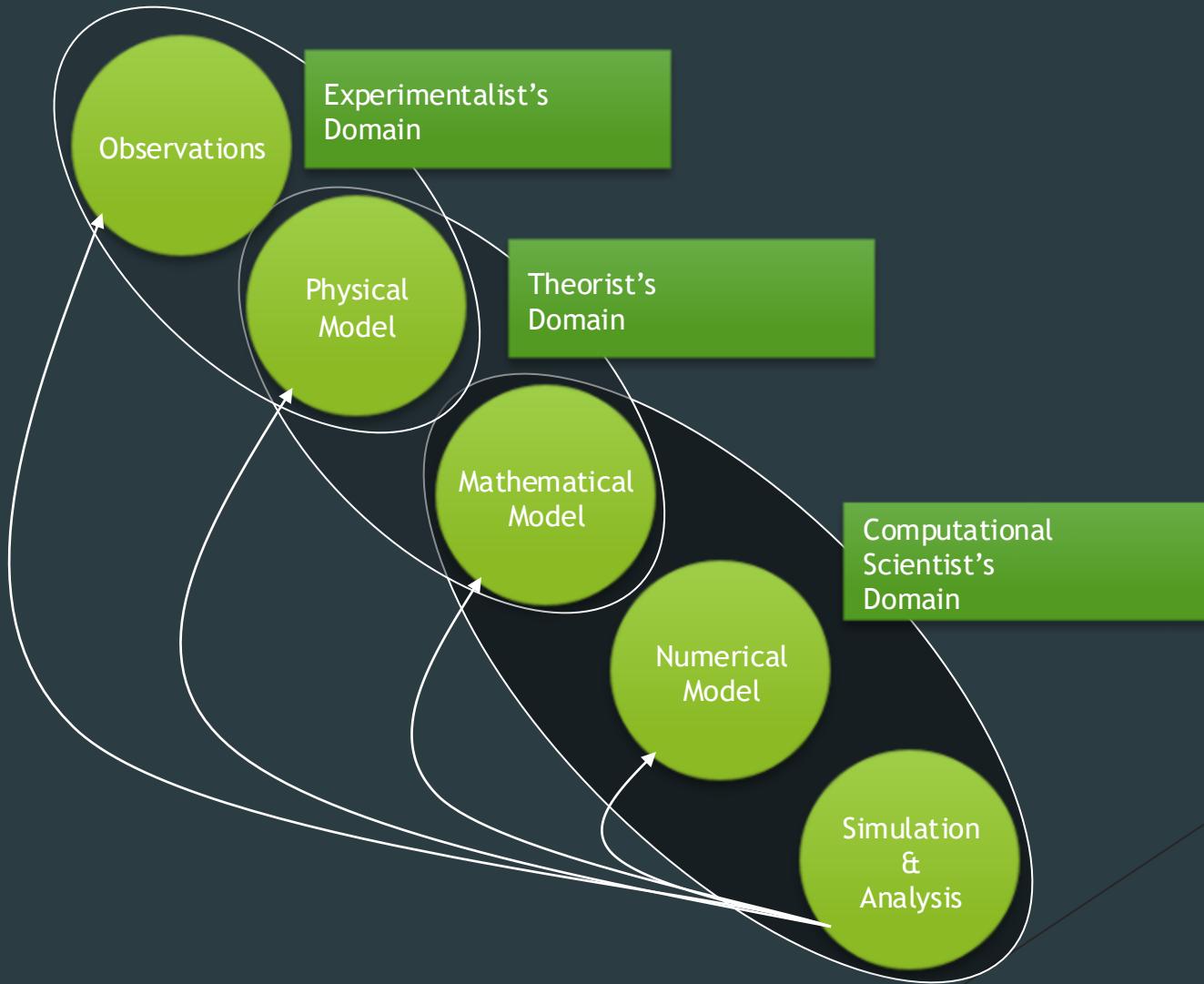
A close-up photograph of a person's hand interacting with a 3D brain scan visualization on a computer monitor. The hand is pointing at a specific area of the brain model, which is overlaid on a grid of 2D MRI slices. The background shows more of the 3D brain model and some numerical values like '119' and '106'.

Scientific Visualization and Virtual Reality

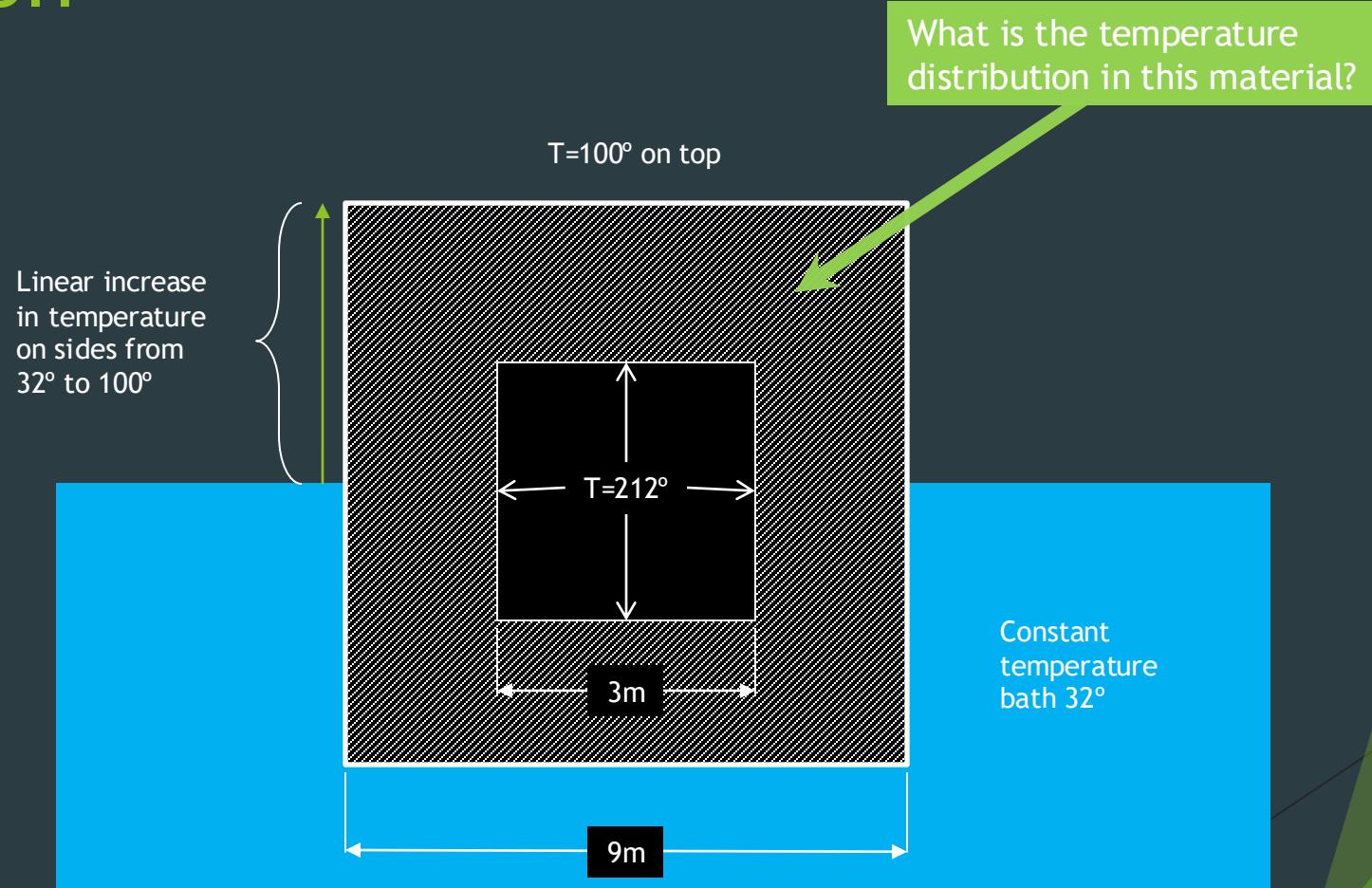
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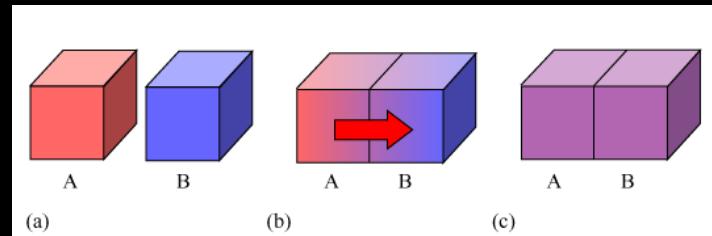
Computational simulation



Example physics problem: temperature distribution



Physical models



(a) Two bodies A and B are of different temperatures, the temperature of A is higher than that of B. (b) When they are in contact, heat is transferred from A to B. (c) Heat transfer will stop when both A and B reach the final temperature

the temperature
ion in this material?



Mathematical model

- ▶ Laplace equation:

$$\nabla^2 c = 0$$

- ▶ Two-dimensional Laplace equation:

$$\nabla^2 c = \frac{\partial^2 c}{\partial x^2} + \frac{\partial^2 c}{\partial y^2} = 0$$

Mathematical model

Solving the Laplace equation using finite differencing
($c_{i,j}$ is the temperature at lattice site (i,j)):

$$\frac{\partial c}{\partial x} \equiv c_{i+1,j} - c_{i,j}$$

$$\frac{\partial c}{\partial y} \equiv c_{i,j+1} - c_{i,j}$$

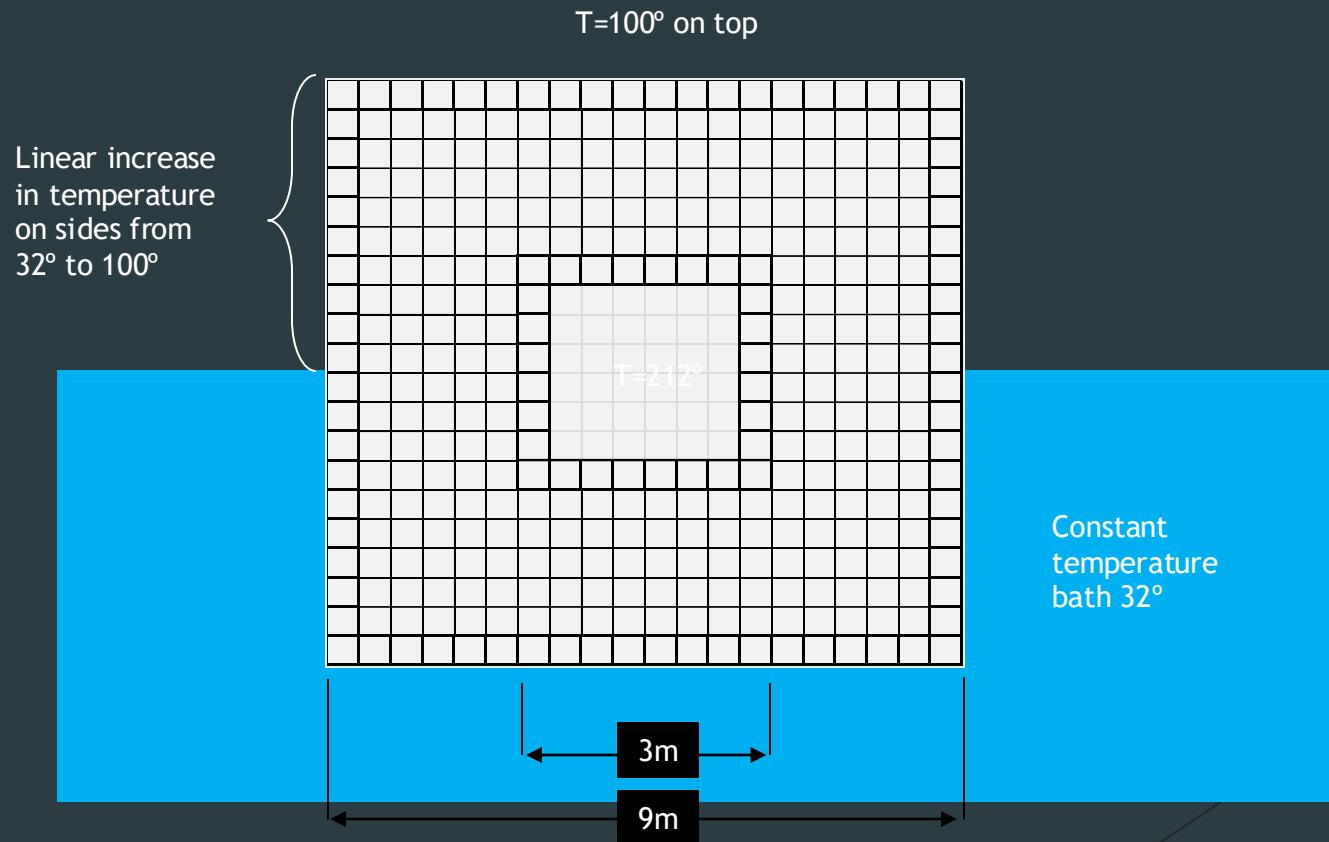
$$\frac{\partial^2 c}{\partial x^2} \equiv (c_{i,j} - c_{i-1,j}) - (c_{i+1,j} - c_{i,j})$$

$$\frac{\partial^2 c}{\partial y^2} \equiv (c_{i,j} - c_{i,j-1}) - (c_{i,j+1} - c_{i,j})$$

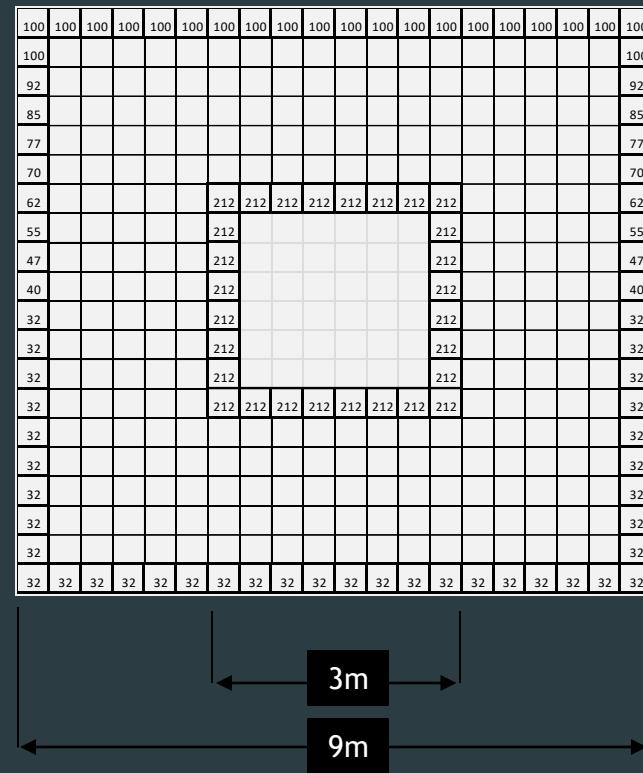
$$\frac{\partial^2 c}{\partial x^2} + \frac{\partial^2 c}{\partial y^2} = 0 \equiv 4c_{i,j} - c_{i-1,j} - c_{i,j-1} - c_{i+1,j} - c_{i,j+1} = 0$$

	$c_{i,j-1}$	
$c_{i-1,j}$	$c_{i,j}$	$c_{i+1,j}$
	$c_{i,j+1}$	

Numerical model



Numerical model



Source: Neal Wagner, Univ. of Texas San Antonio
(original source: <http://www.cs.utsa.edu/~wagner/CS2073/distill/dsgif.html>)

Numerical model

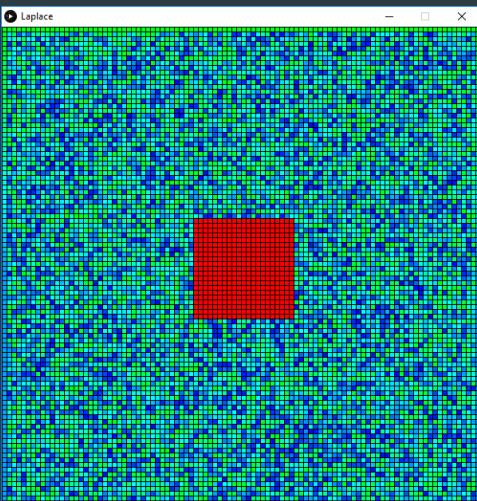
Laplace equation - Jacobi update procedure where $c_{i,j}^n$ is the old value, $c_{i,j}^{n+1}$ is the new value:

```
do {  
     $c_{i,j}^{n+1} = \frac{1}{4}(c_{i-1,j}^n + c_{i,j-1}^n + c_{i+1,j}^n + c_{i,j+1}^n)$   
} while ( $|c_{i,j}^n - c_{i,j}^{n+1}| > tolerance$ )
```

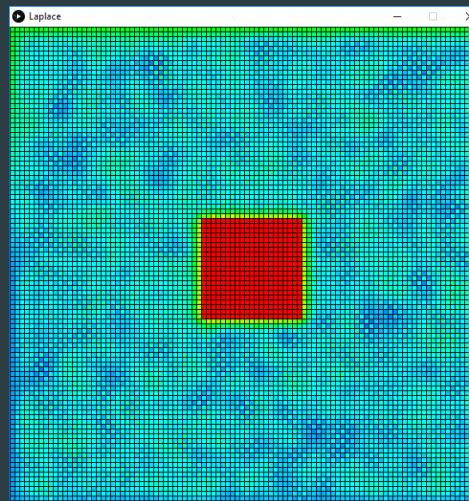
Simulation pseudocode

```
initialize all lattice sites
set boundary conditions
do {
    δ = 0
    for i=1 to IMAX {
        for j=1 to JMAX {
            west_neighbour = (i==1) ? cnIMAX,j : cni-1,j
            east_neighbour = (i==IMAX) ? cn1,j : cni+1,j
            north_neighbour = (j==JMAX) ? cni,1 : cni,j+1
            south_neighbour = (j==1) ? cni,JMAX : cni,j-1
            cn+1i,j=%(west_neighbour + east_neighbour +
                           north_neighbour + south_neighbour)
            if (cni,j - cn+1i,j > tolerance) δ = cni,j - cn+1i,j
        }
    }
} while (δ > tolerance)
```

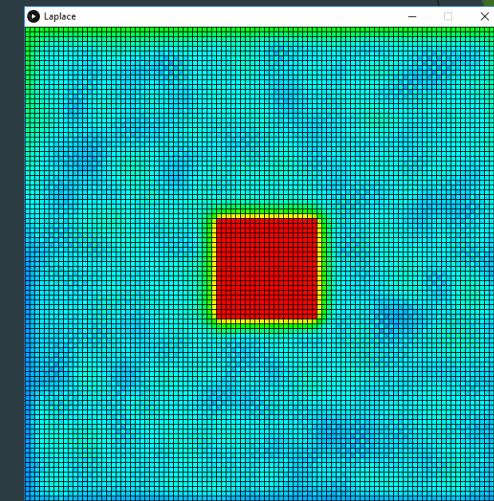
Simulate!



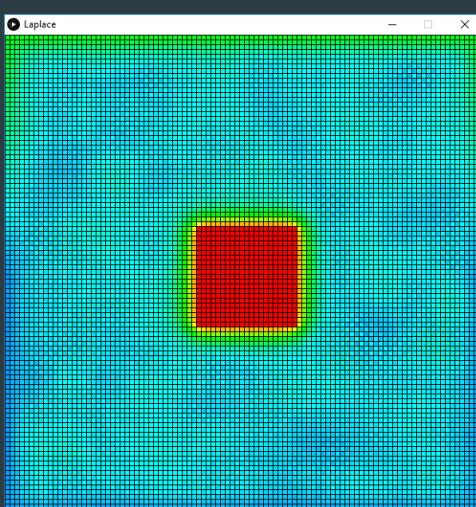
$t=0$ ($\delta \approx 100$)



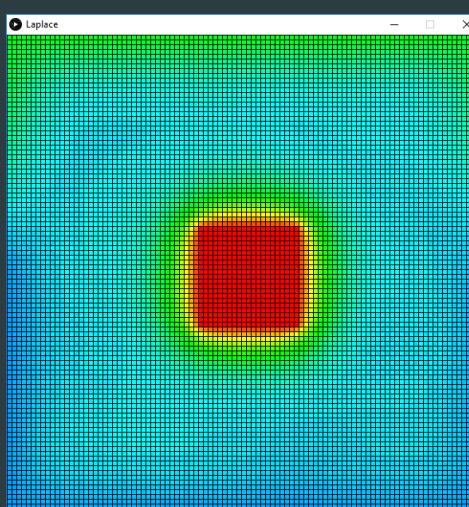
$t=5$ ($\delta \approx 50$)



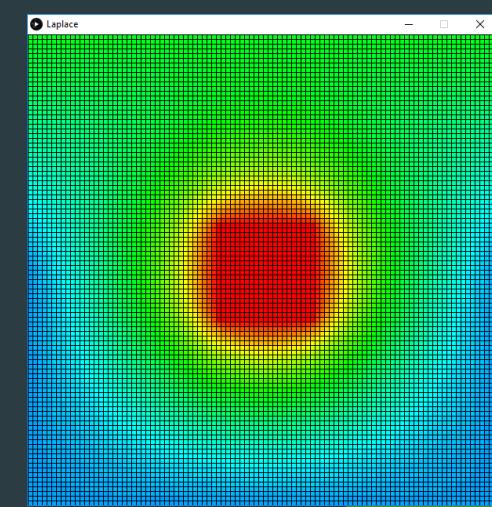
$t=10$ ($\delta \approx 40$)



$t=20$ ($\delta \approx 30$)



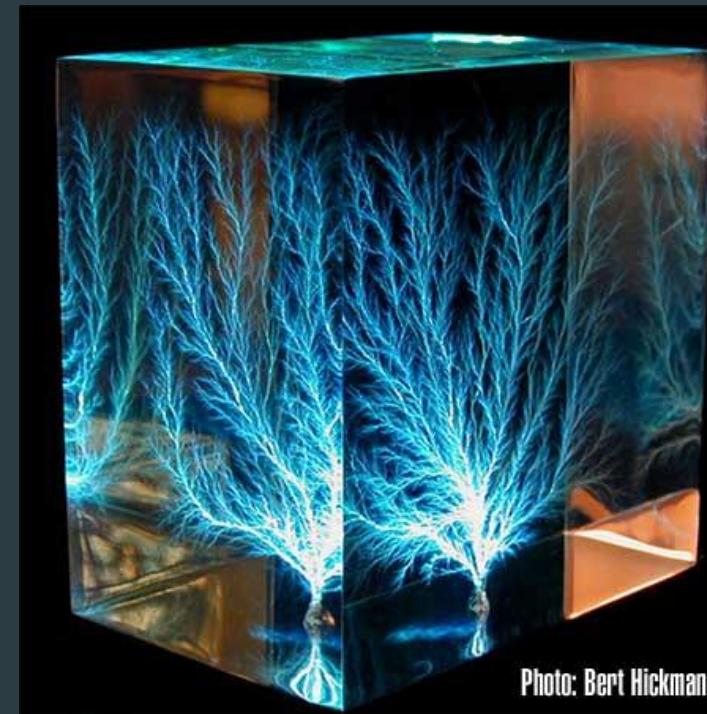
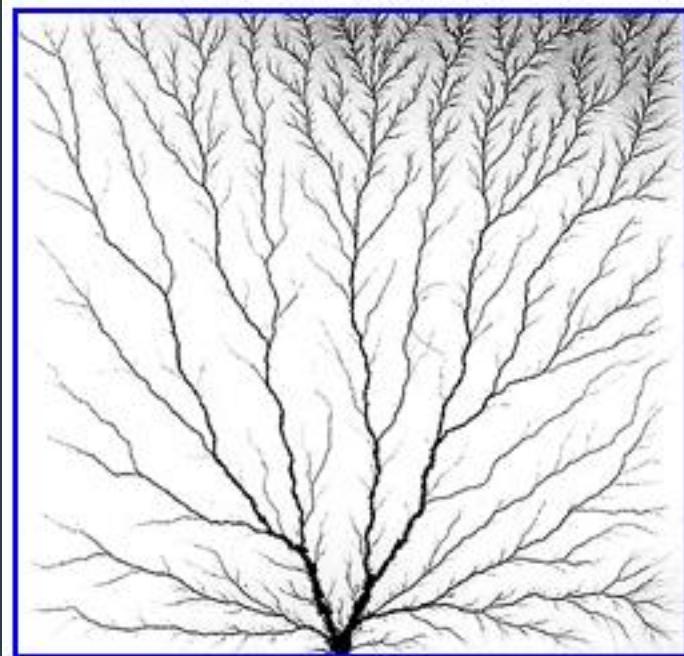
$t=100$ ($\delta \approx 17$)



$t=1000$ ($\delta \approx 9$)

Example: Abiotic growth and form; Lichtenberg patterns

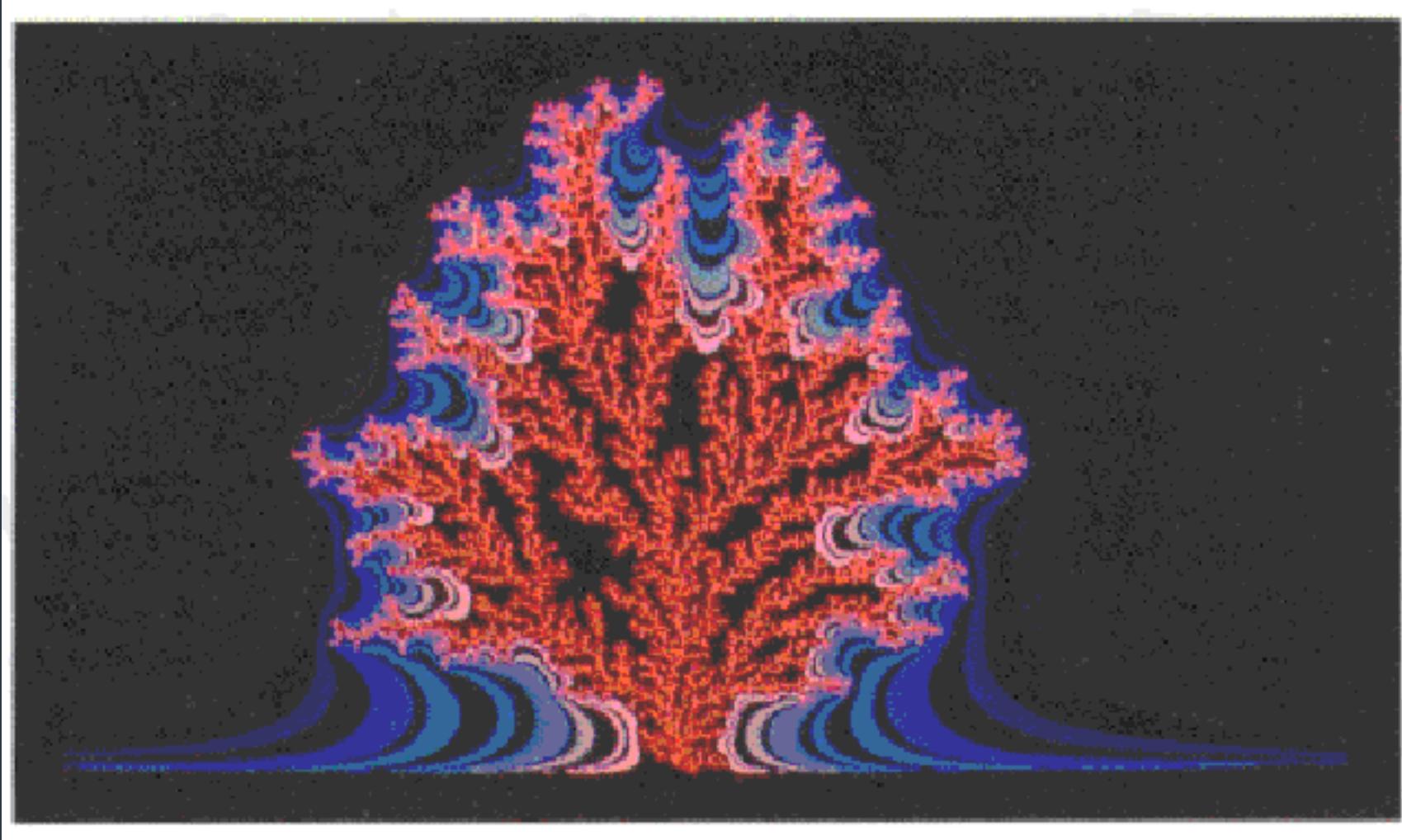
- ▶ “Dielectric breakdown”
- ▶ First created by Lichtenberg,
18th century



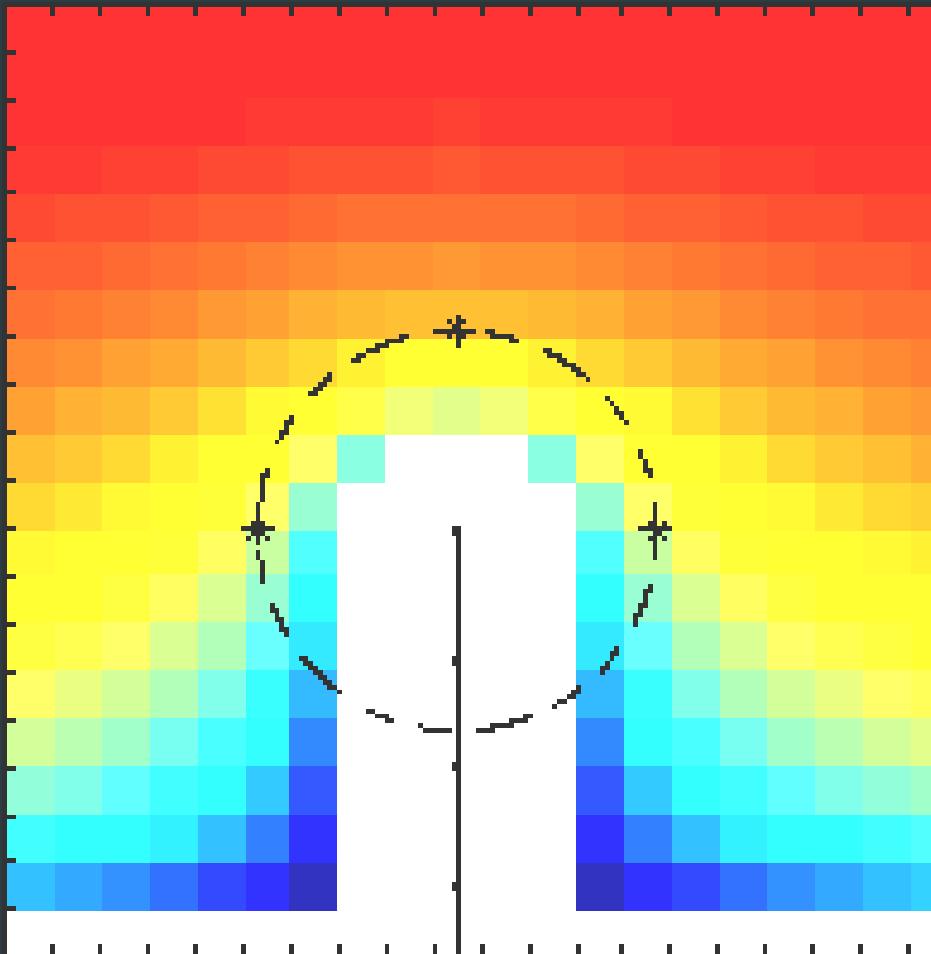
Example: Abiotic growth and form;
Lichtenberg patterns



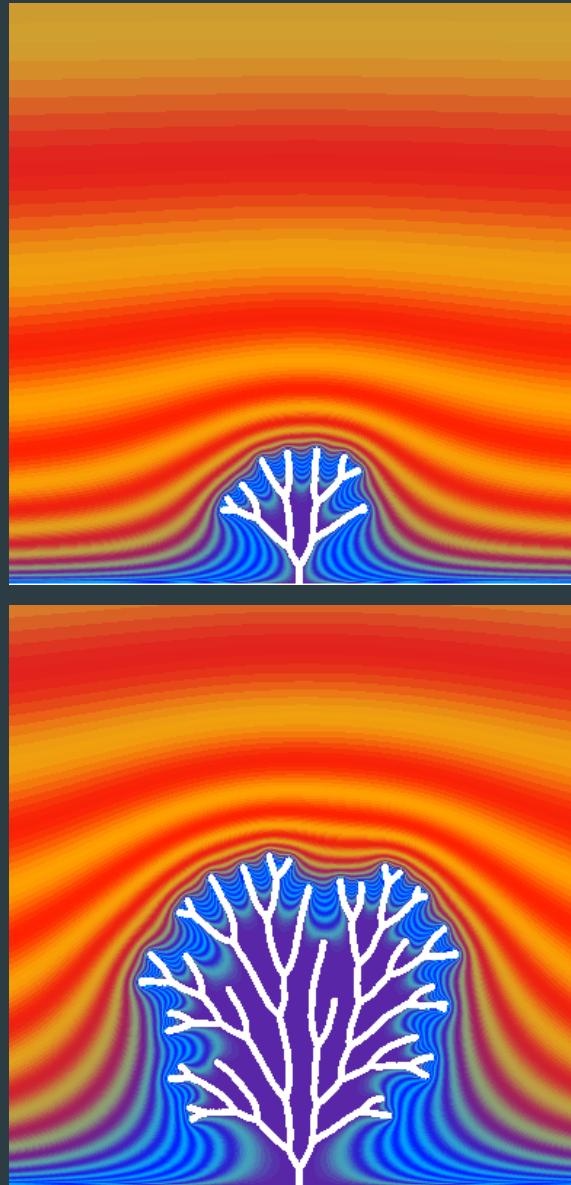
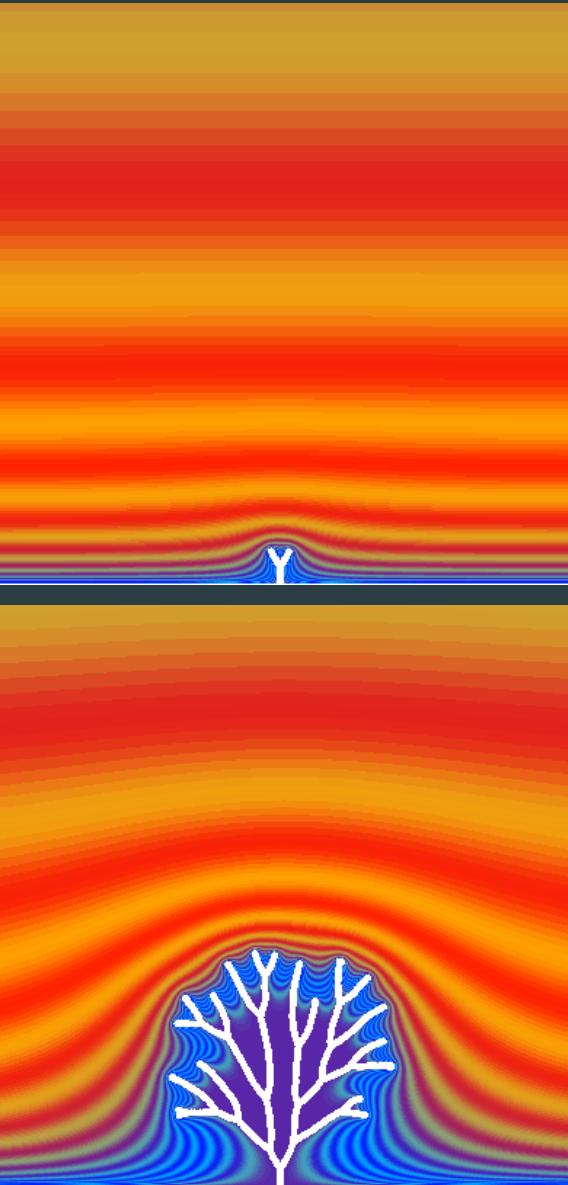
Laplacian growth patterns



Laplacian growth patterns

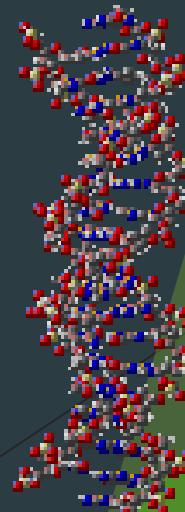
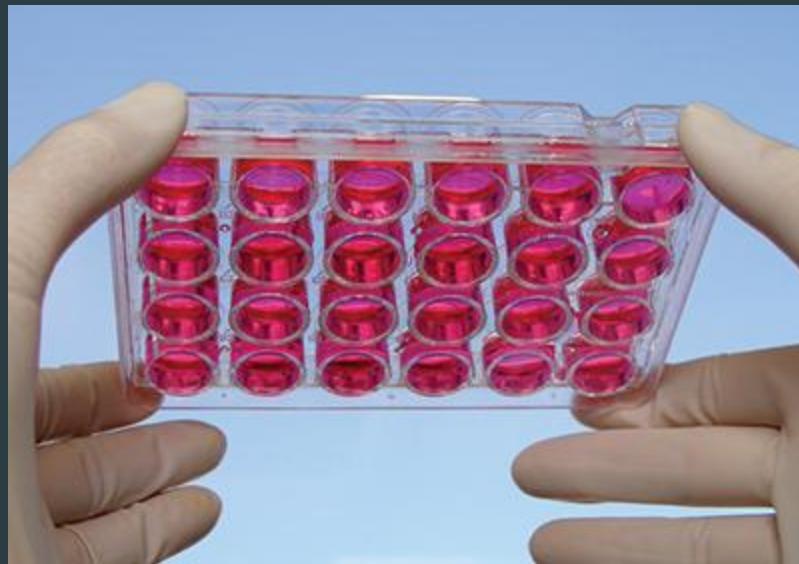
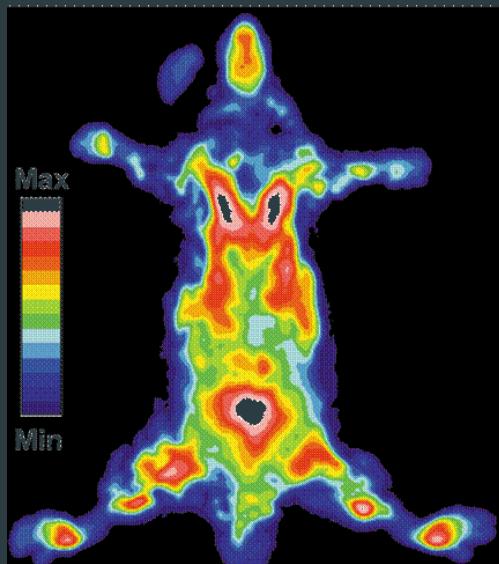


Laplacian growth patterns



In vivo, in vitro, in silico

- ▶ *in vivo*: in the living organism
- ▶ *in vitro*: in an artificial environment outside the living organism
- ▶ *in silico*: in a computer simulated model



Science in silico

- ▶ SEED Magazine, April 19, 2007
Source: http://seedmagazine.com/content/article/science_in_silico/
- ▶ (also available [on YouTube](#))



Science in Silico

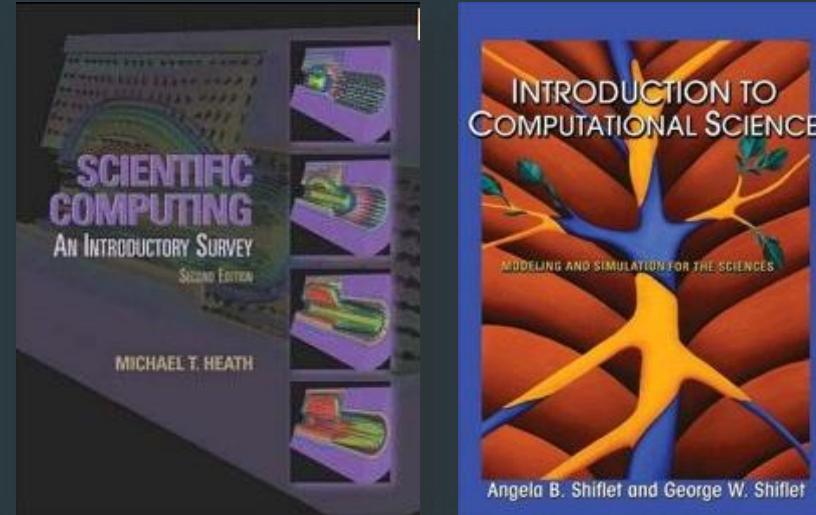


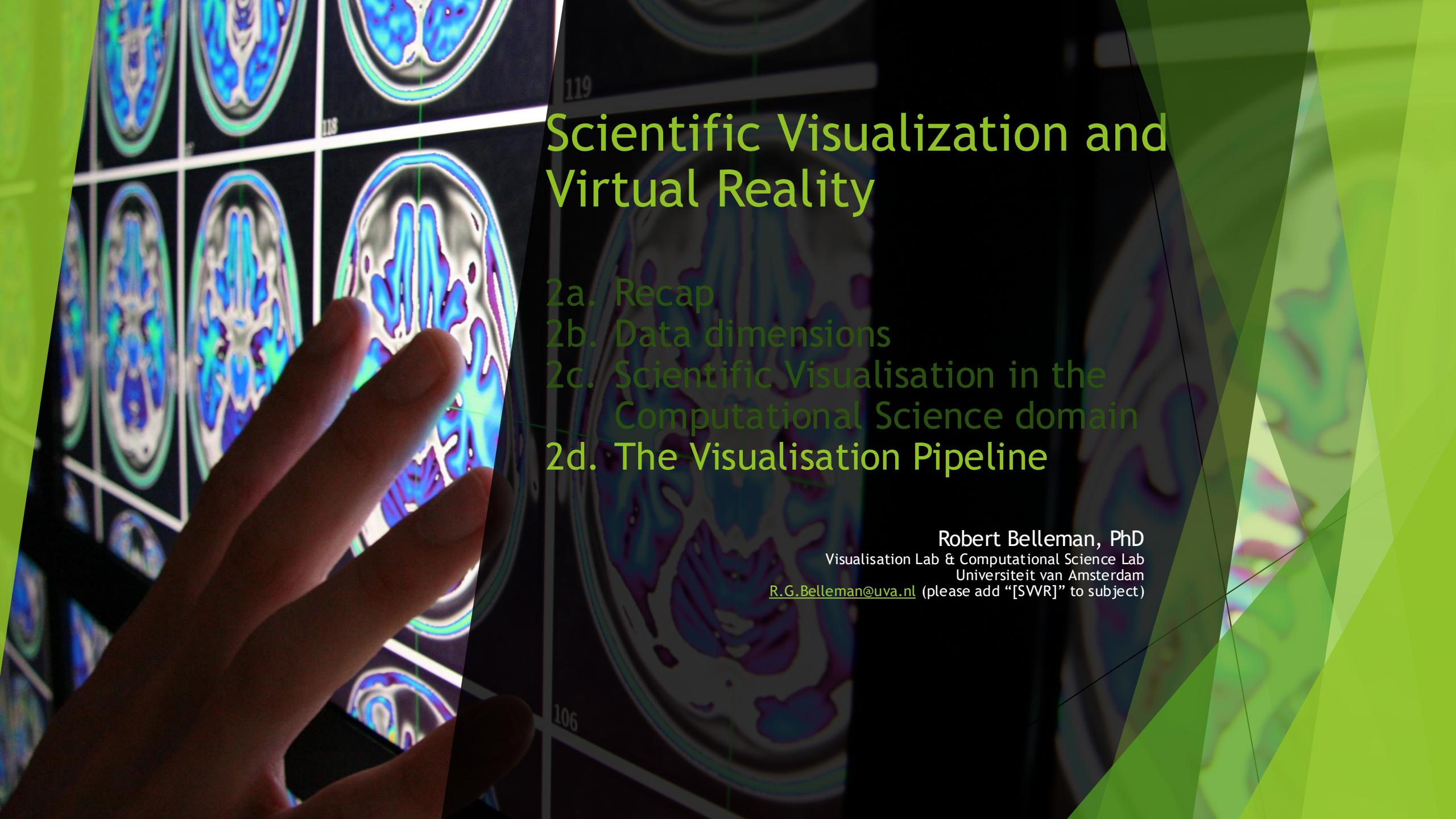
<https://www.youtube.com/watch?v=1FvRSJ9W734&t=1s>

If you want more

Book suggestions:

- ▶ Michael T. Heath, *Scientific Computing, an introduction survey*
- ▶ A.B. Shiflet and G.W. Shiflet, *Introduction to Computational Science: Modeling and Simulation for the Sciences*,



A close-up photograph of a person's hand interacting with a 3D brain scan visualization on a computer monitor. The hand is pointing at a specific area of the brain model, which is overlaid on a series of axial MRI slices. The brain model is color-coded, showing different regions in shades of blue, green, and yellow. The monitor displays a grid of these slices, with the number '119' visible in the top right corner. The background is dark, making the bright colors of the brain model stand out.

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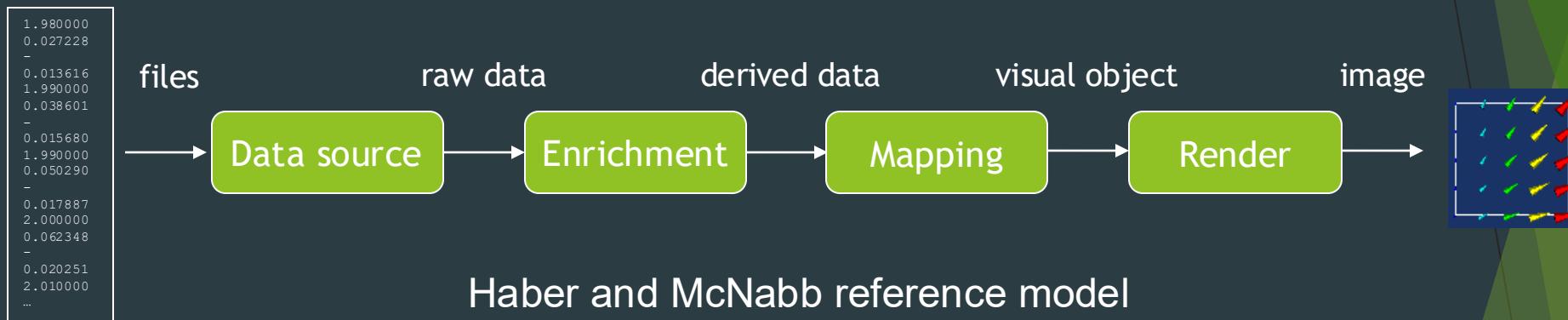
The visualization pipeline



Haber and McNabb reference model

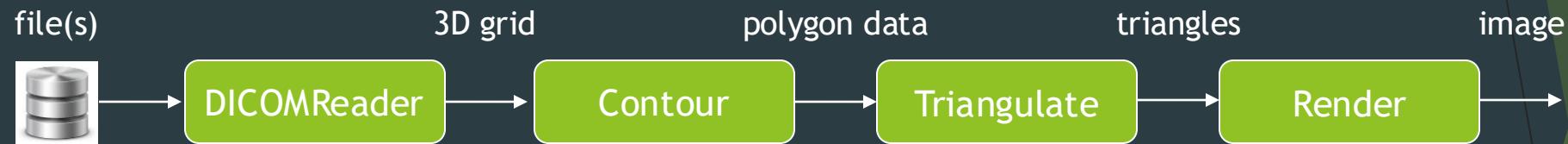
Haber, R.B. and McNabb, D.A. *Visualization idioms: A conceptual model for scientific visualization systems*. In *Visualization in Scientific Computing*, pages 74-93. IEEE Computer Society Press. 1990

The visualization pipeline

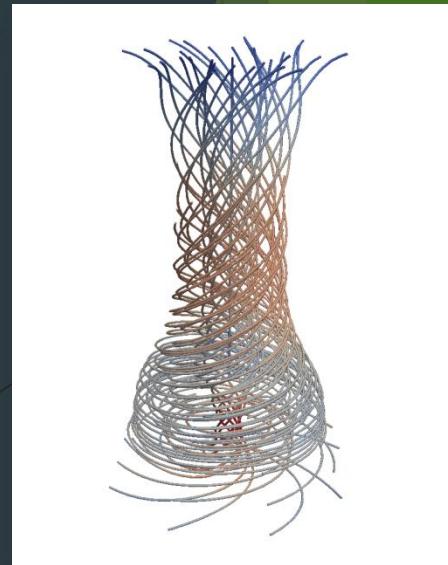
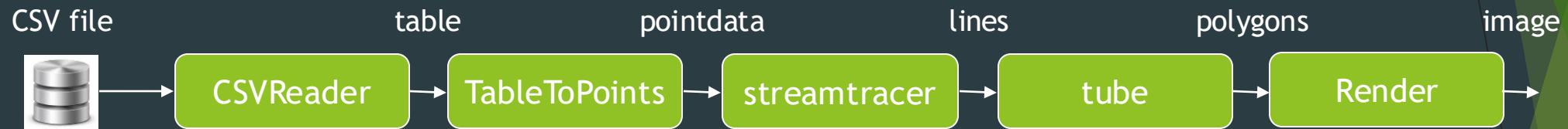


Haber, R.B. and McNabb, D.A. *Visualization idioms: A conceptual model for scientific visualization systems*. In *Visualization in Scientific Computing*, pages 74-93. IEEE Computer Society Press. 1990

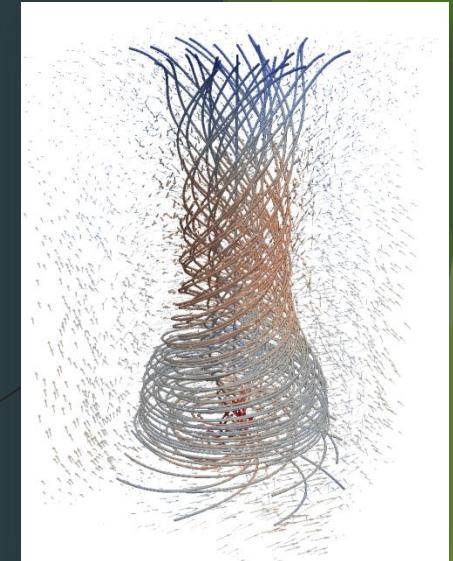
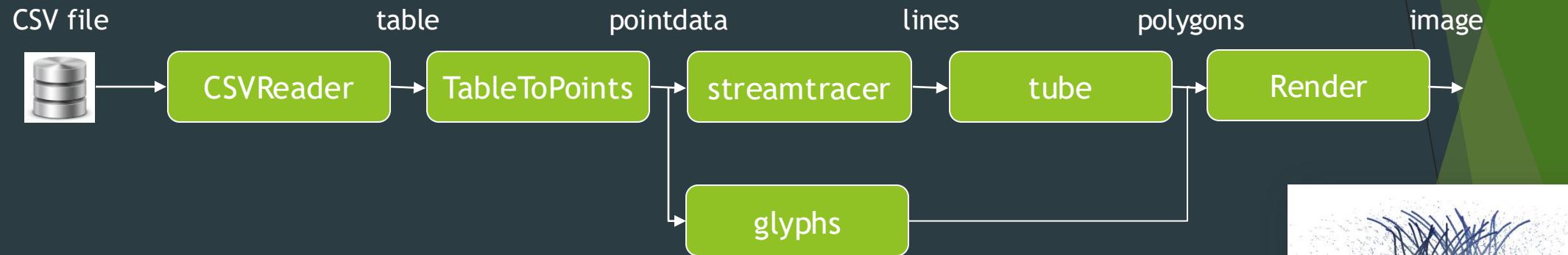
Example: extracting a contour from medical data



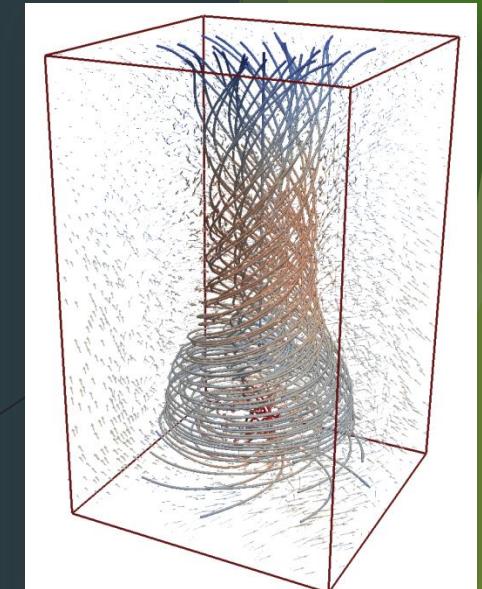
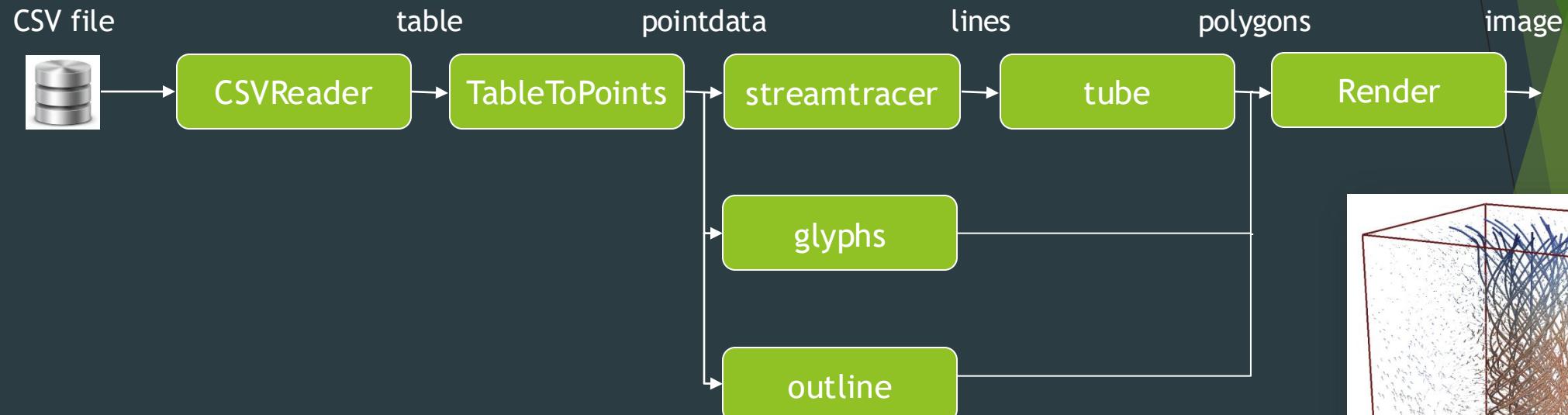
Example: visualizing flow data with streamlines



Example: visualizing flow data with streamlines



Example: visualizing flow data with streamlines



Pipeline creation

- ▶ The components in a pipeline are called “filters”
- ▶ Filters are connected together to form a “visualization pipeline” or “dataflow network”
- ▶ The input port of a filter may only be connected to the output port(s) of (an)other filter(s) if the port types are “similar”



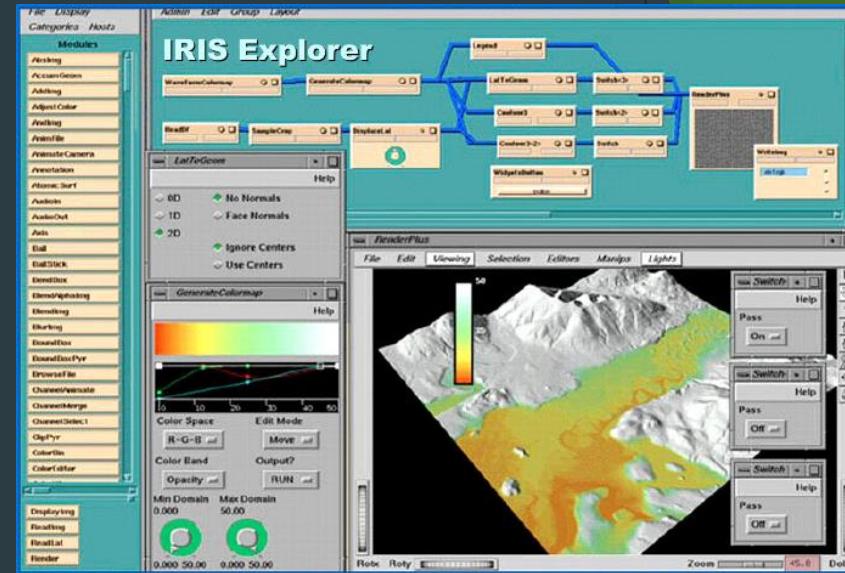
Pipeline behaviour

- ▶ Filters in a pipeline only execute when necessary
 - ▶ When data at the input has changed, or a parameter
- ▶ Data flows downstream, update checks go upstream



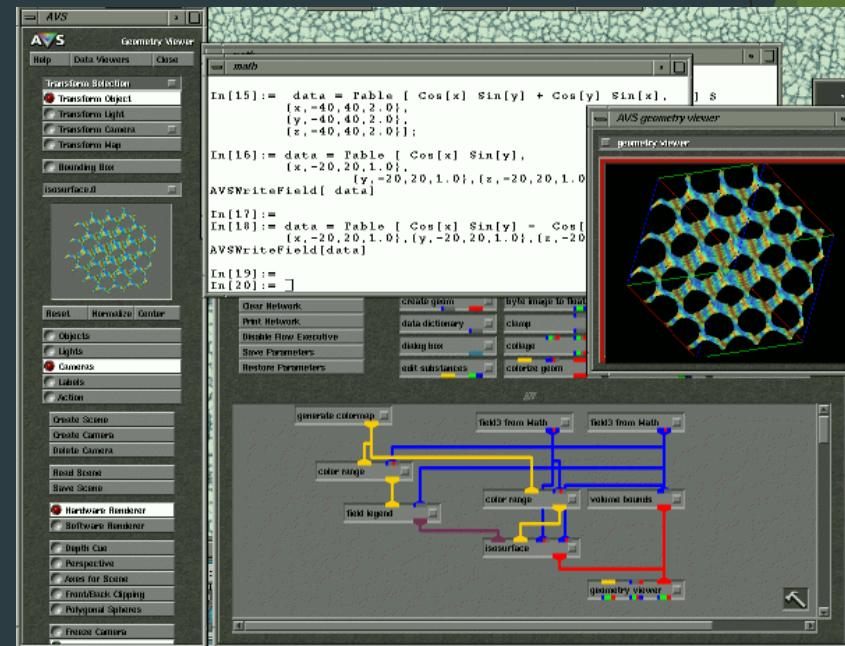
Software

- ▶ Often domain-specific
- ▶ Almost all based on visualization pipeline / dataflow concept



Commercial:

- ▶ AVS (Advanced Visual Systems)
- ▶ OpenDX (now open source?)
- ▶ IRIS Explorer (?)
- ▶ Amira
- ▶ Matlab, Mathematica, IDL
- ▶ Spotfire, Tableau
- ▶ ...



Software

- ▶ Often domain-specific
- ▶ Almost all based on visualization pipeline / dataflow concept

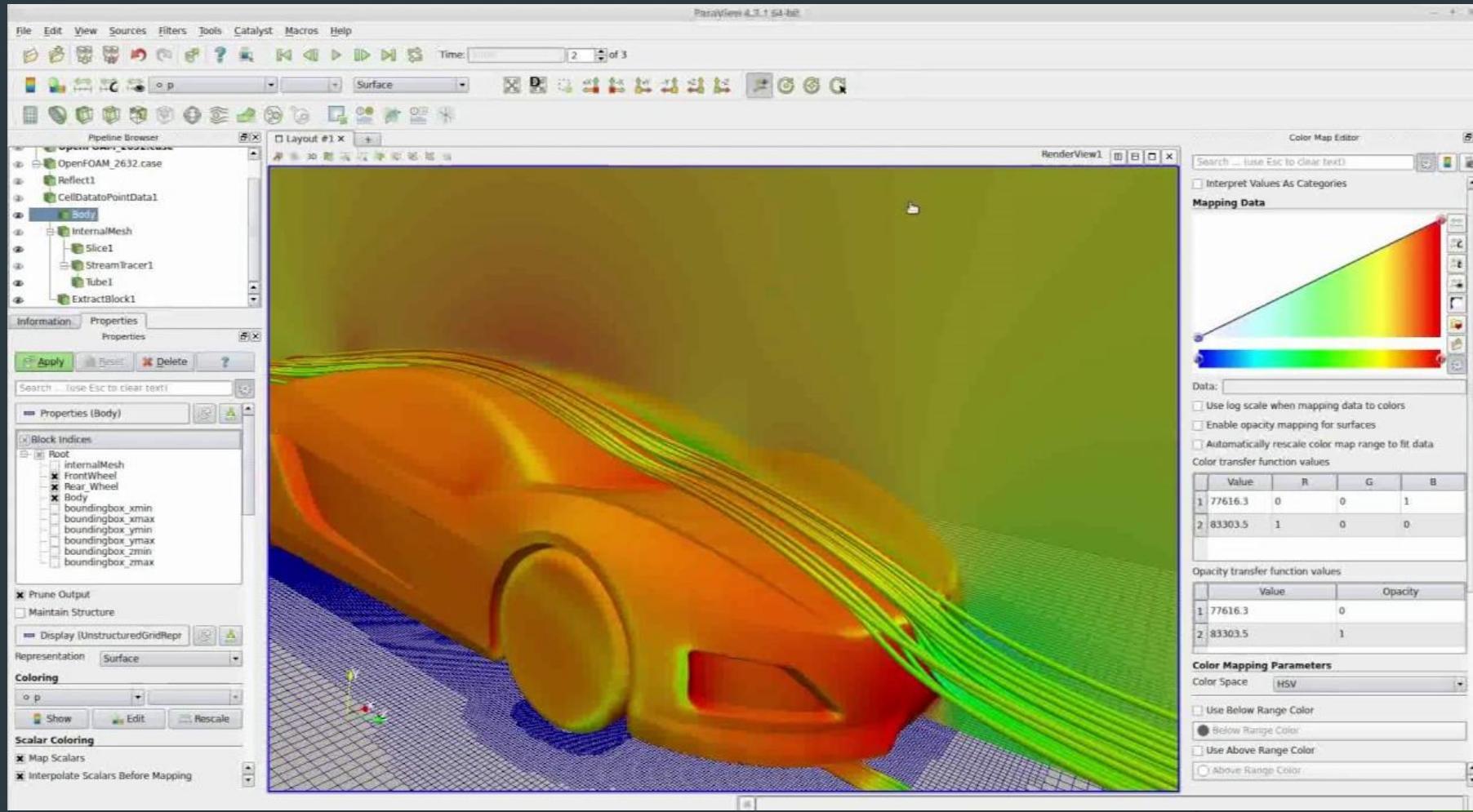
Public domain:

- ▶ ParaView 
- ▶ Visualization toolkit (VTK) 
- ▶ 3D Slicer 
- ▶ VolView 
- ▶ MayaVI, VisIt, DeVIDE, MeVisLab,
SCIRun, Gephi, Cytoscape, R, ...



ParaView: install v5.13 before next Wednesday!

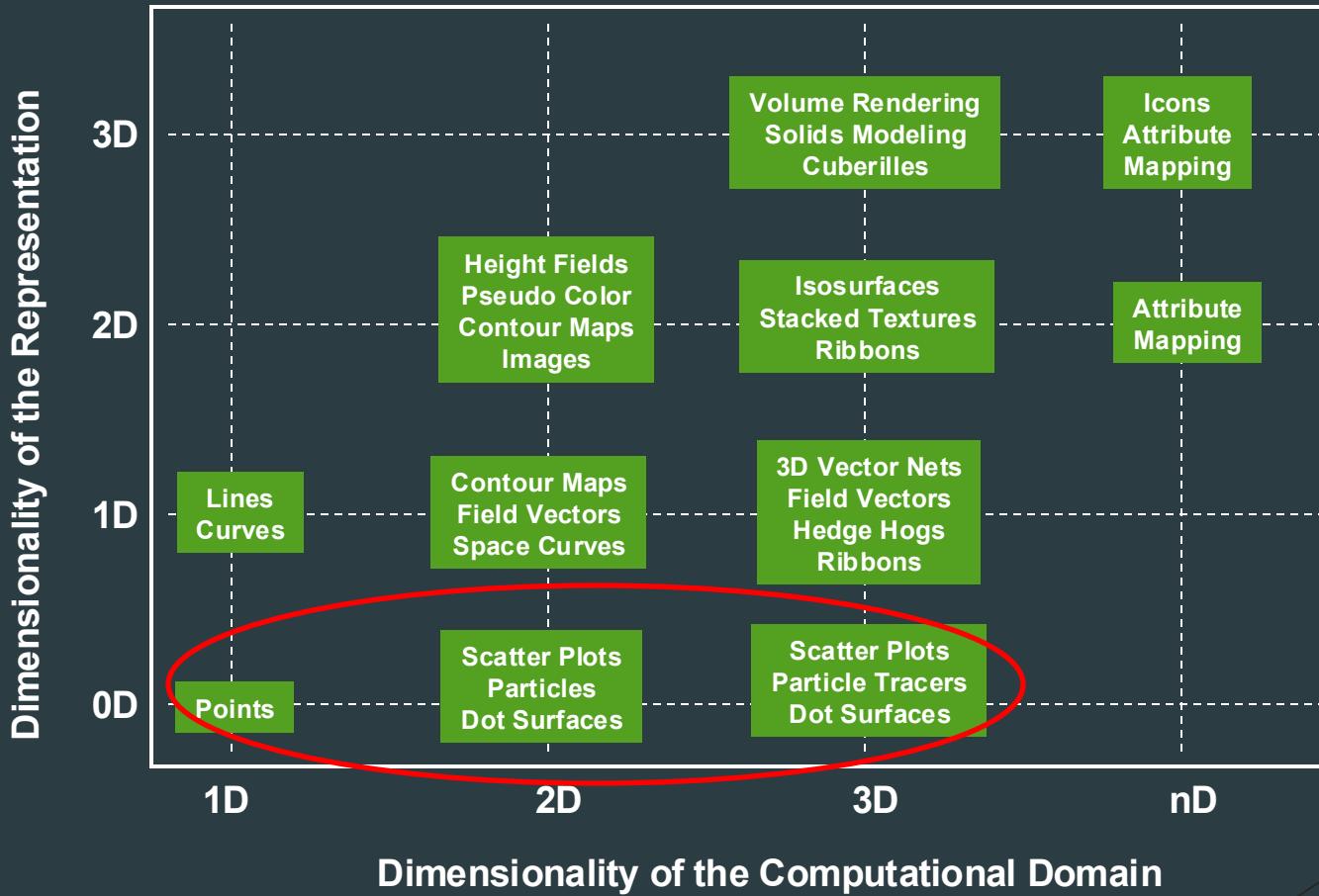
Go to <https://www.paraview.org/download/>





Scientific Visualisation I

Visualization mapping space

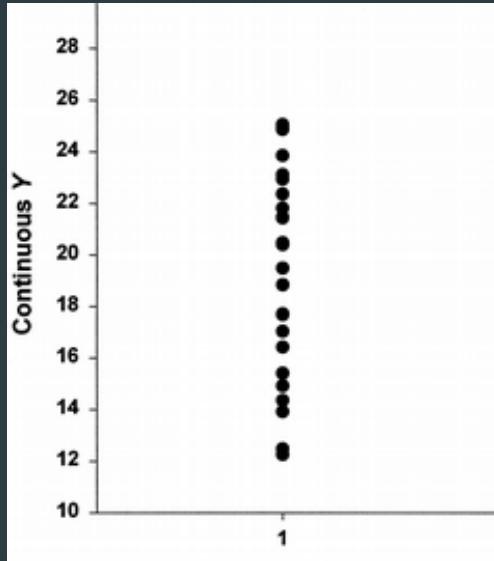


Based on: Craig Upson, *Volumetric Visualization Techniques*, in "State of the Art in Computer Graphics: Visualization and Modeling", David F. Rogers Rae A. Earnshaw (eds.), Springer-Verlag, 1991, pp. 313-350

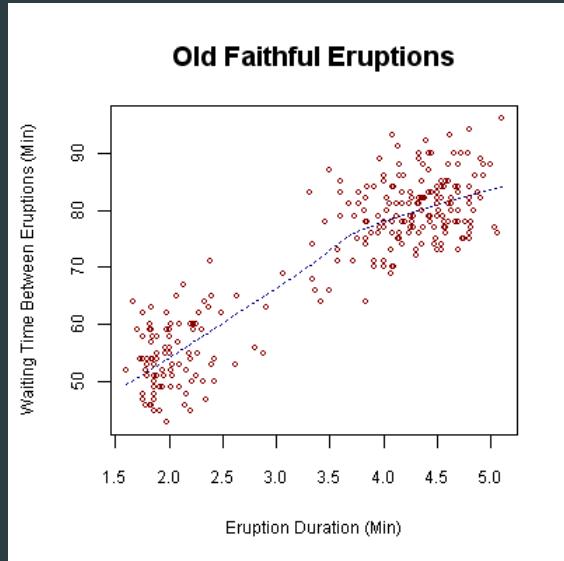
Zero-dimensional primitives

- ▶ Dots •
- ▶ Points ○
- ▶ Markers ◇
- ▶ (Tracked) particles ↗

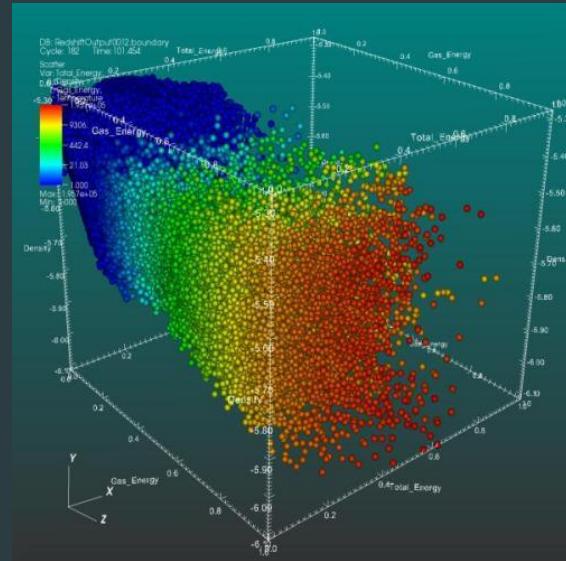
Scatterplot



1D computational domain

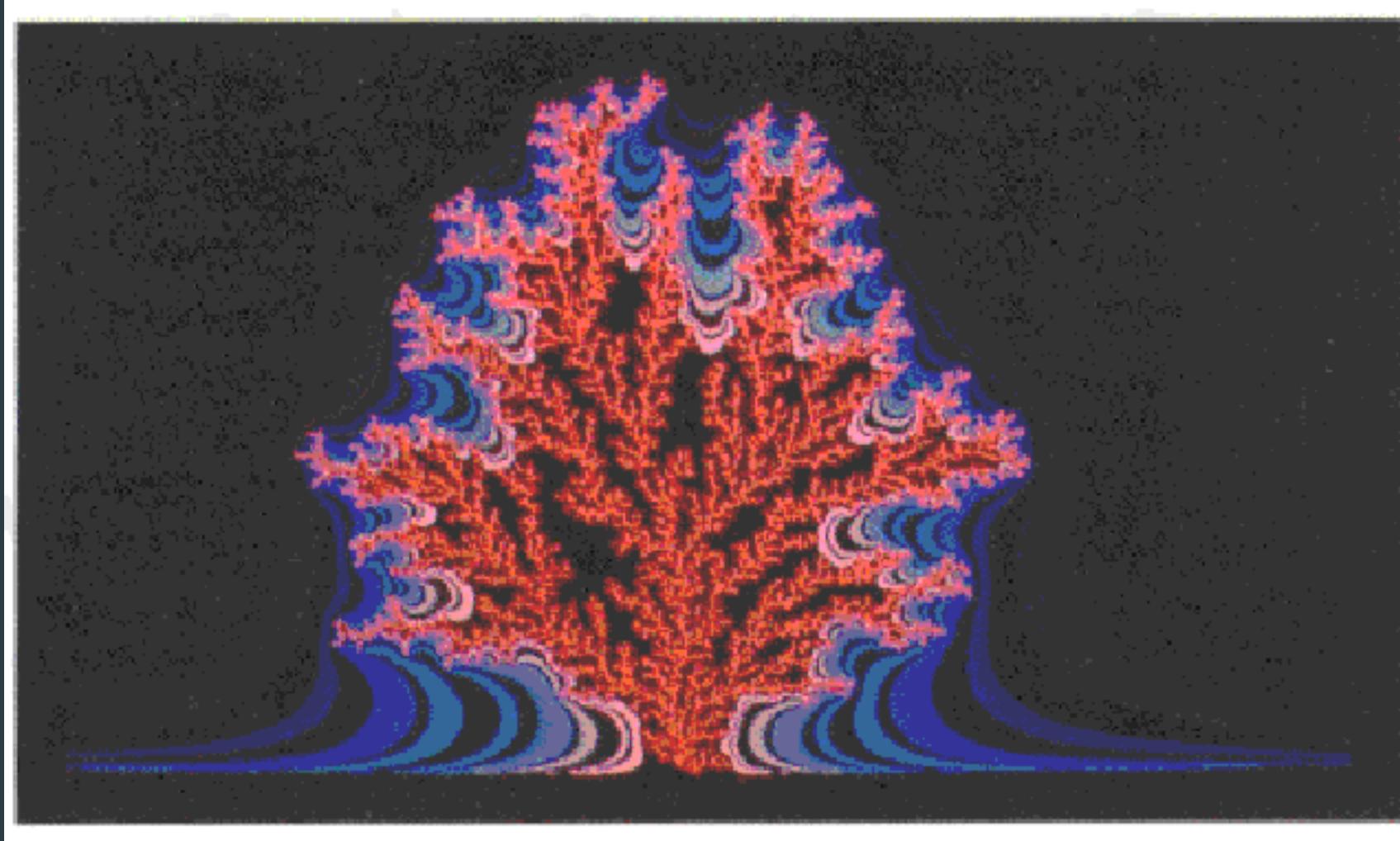


2D computational domain

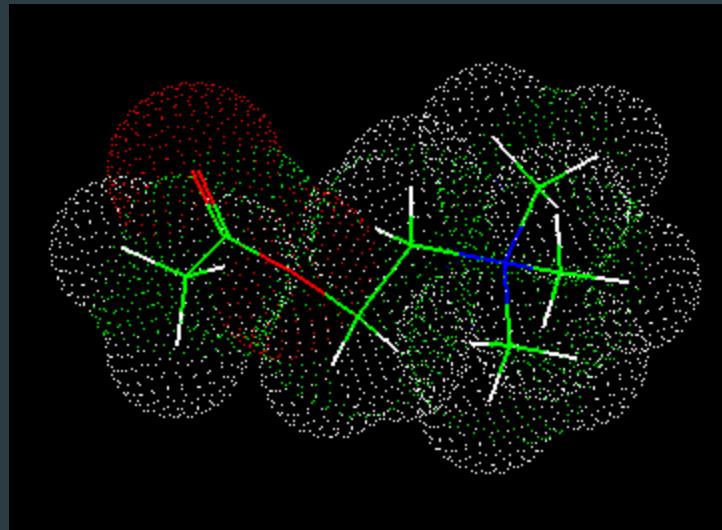


3D computational domain

Dot surface representation of Laplacian growth pattern

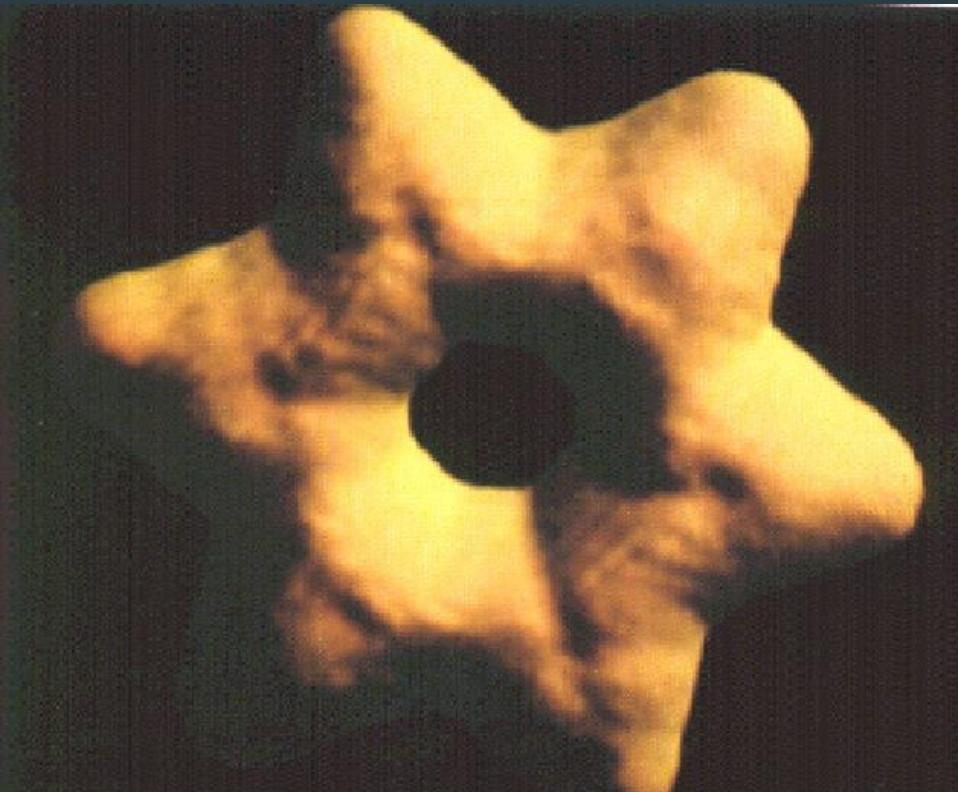


Dot surface representation of a molecule



Physical boundaries of a molecule: Van der Waals surface representation of acetylcholine

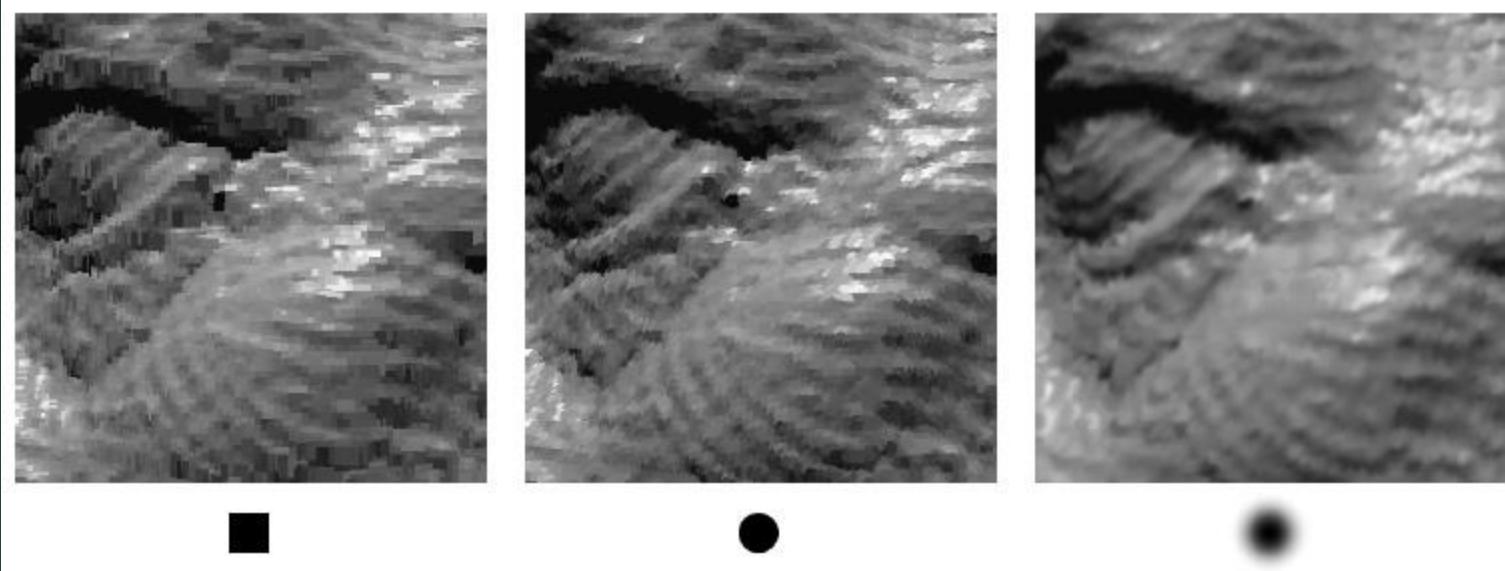
Dot surface representation benzene's electron density function



Craig Upson, *Volumetric Visualization Techniques*, in "State of the Art in Computer Graphics: Visualization and Modeling", David F. Rogers Rae A. Earnshaw (eds.), Springer-Verlag, 1991, pp. 313-350

Point rendering

- ▶ Splatting

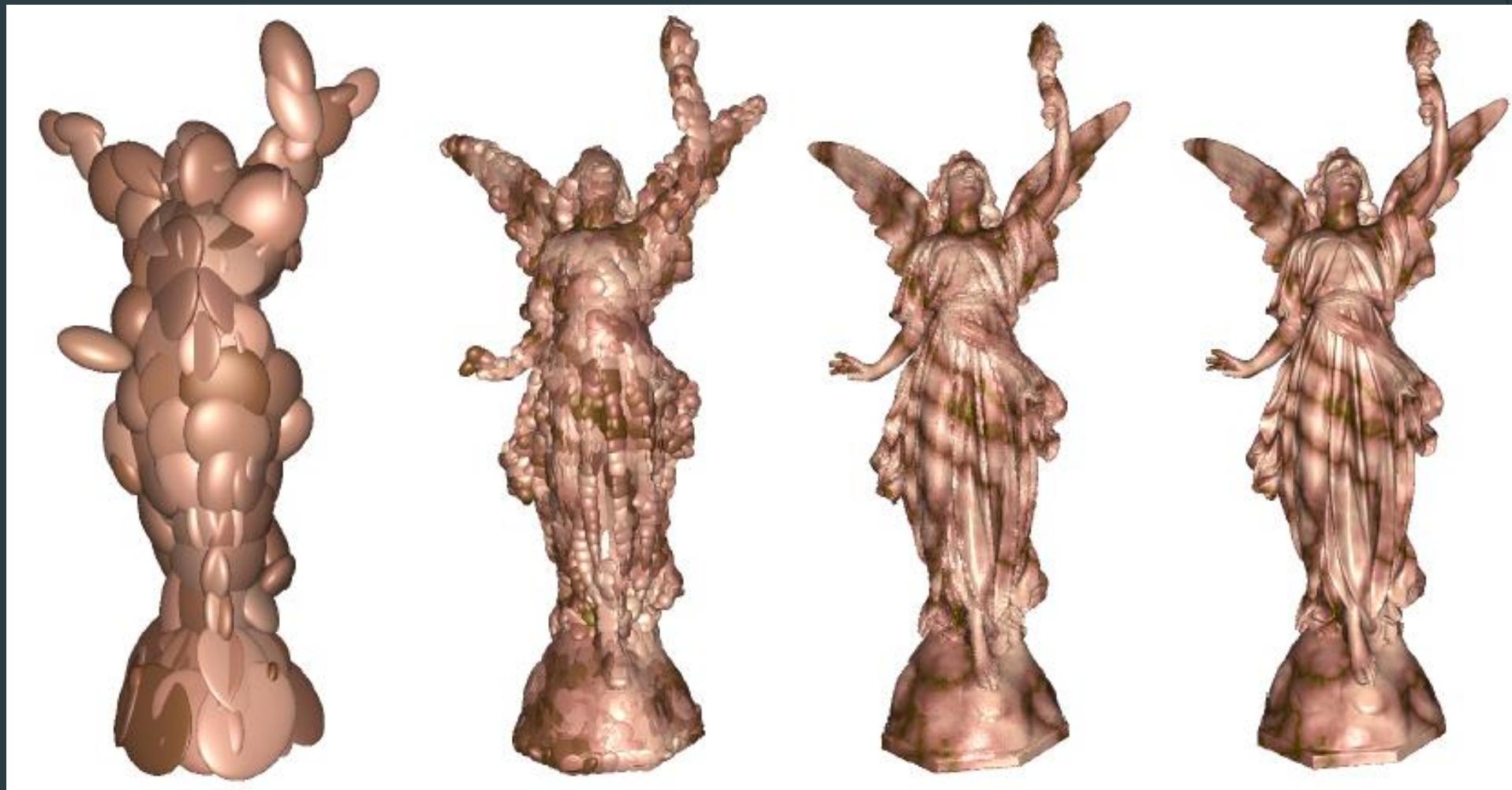


Szymon Rusinkiewicz, Marc Levoy, QSplat: A Multiresolution Point Rendering System for Large Meshes, SIGGRAPH 2000, pp 343 - 352

Point cloud rendering



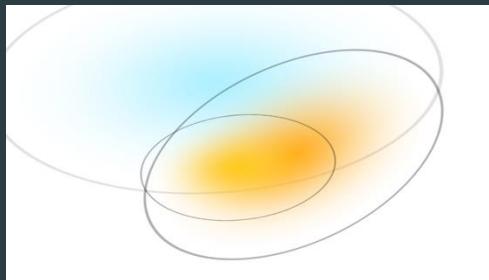
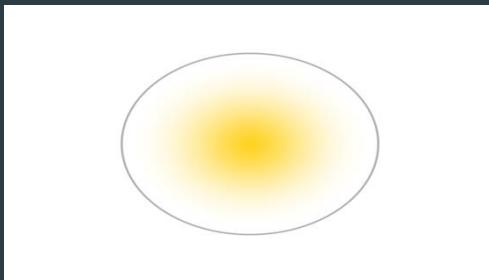
Source: Valentin Kraft, Univ. Bremen ([YouTube](#))



A. Kalaiah and A. Varshney, Statistical Point Geometry, Symposium on Geometry Processing 2003, pp 107 - 115

Gaussian splatting

- ▶ “A method using 3D Gaussians for scene representation and optimized rendering allows high-quality, real-time novel-view synthesis at 1080p resolution”
 - **Position:** where it's located (XYZ)
 - **Covariance:** how it's stretched/scaled (3x3 matrix)
 - **Color:** what color it is (RGB)
 - **Alpha:** how transparent it is (α)



Source: [Kerbl et al. 3D Gaussian Splatting for Real-Time Radiance Field Rendering](#).

See also: “[What is 3D Gaussian splatting?](#)” on YouTube.



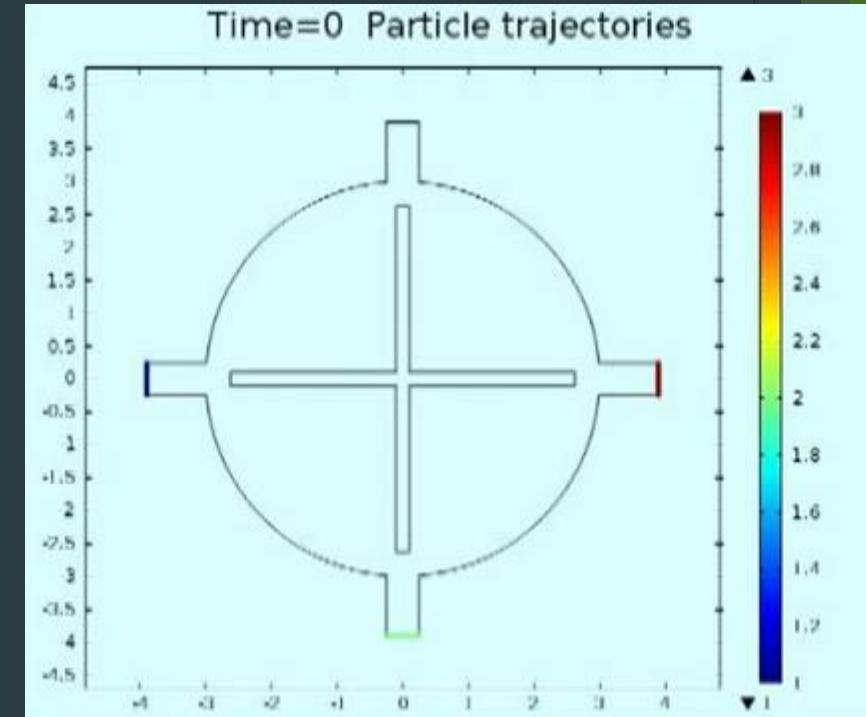
Gaussian splat using 7 million gaussians, with opacity



Gaussian splat using 7 million gaussians, without opacity

Particle tracers: advection

- ▶ Data domain: unsteady rectilinear grid of vectors
- ▶ Associated with each particle:
 - ▶ Location (x,y,z)
 - ▶ Colour triplet (R,G,B)
 - ▶ Transparency
 - ▶ Size
 - ▶ Lifespan



Y. Yang, Particle tracing in a Micromixer
Source: <https://www.youtube.com/watch?v=GPyHT-r0XpI>

Particle advection

- ▶ Solve equation of motion:

$$V = dP/dt$$

where:

V is the vector velocity (u, v, w)
 P is the particle location (x, y, z)

- ▶ Euler approximation:

$$\begin{aligned}x(t + dt) &= x(t) + u(x(t), y(t), z(t)) dt \\y(t + dt) &= y(t) + v(x(t), y(t), z(t)) dt \\z(t + dt) &= z(t) + w(x(t), y(t), z(t)) dt\end{aligned}$$

Interpolation in a cell

$$A_1 = (x - x(i)) (y - y(j))$$

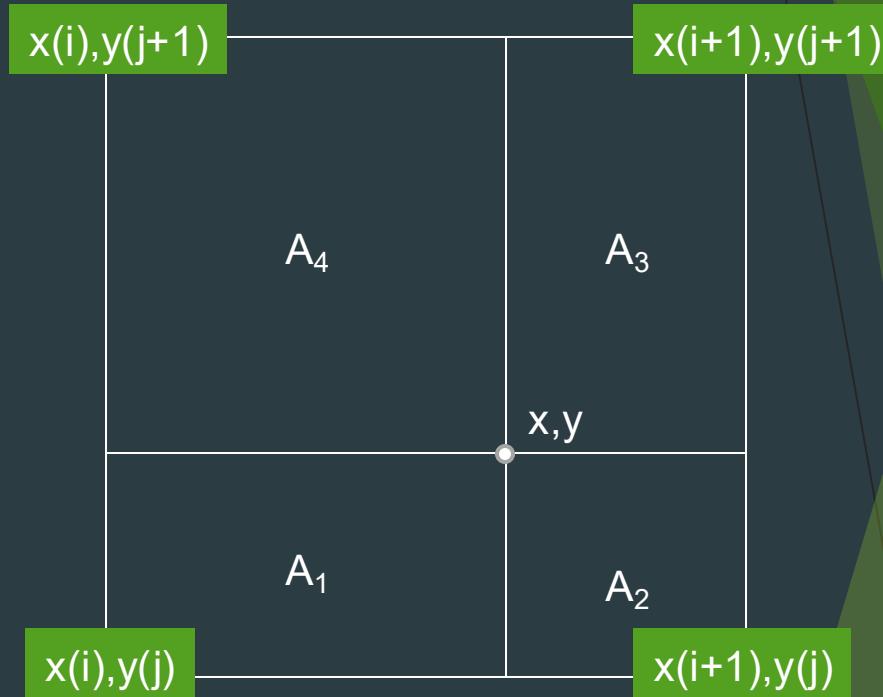
$$A_2 = (x(i+1) - x) (y - y(j))$$

$$A_3 = (x(i+1) - x) (y(j+1) - y)$$

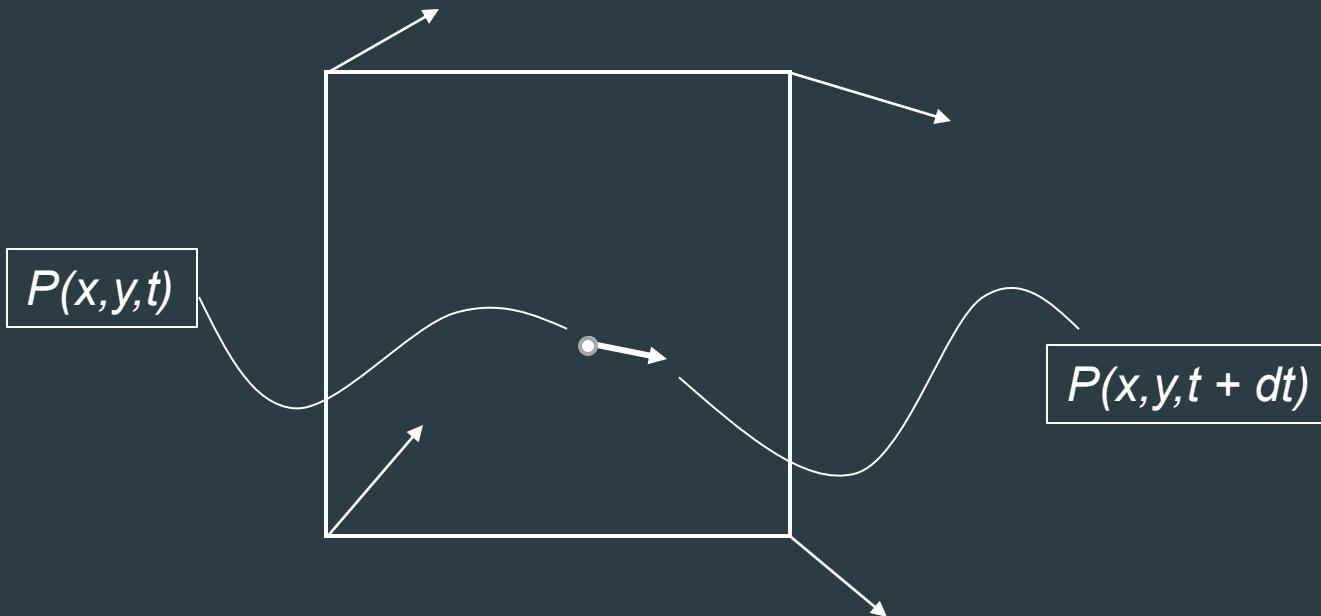
$$A_4 = (x - x(i)) (y(j+1) - y)$$

Interpolated velocity:

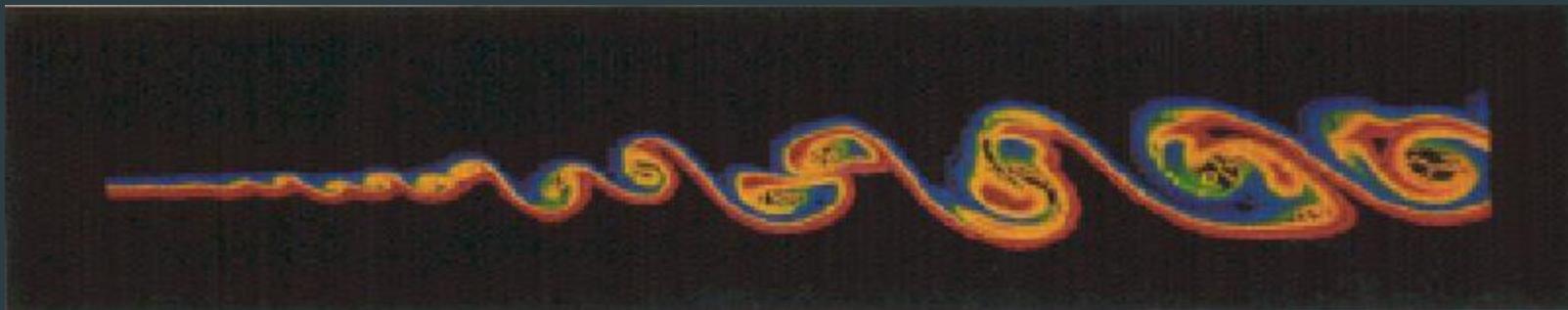
$$u(x,y) = A_1 u(x(i+1),y(j+1)) + \\ A_2 u(x(i),y(j+1)) + \\ A_3 u(x(i),y(j)) + \\ A_4 u(x(i+1),y(j))$$



Particle motion



Large scale structure in a turbulent shear layer



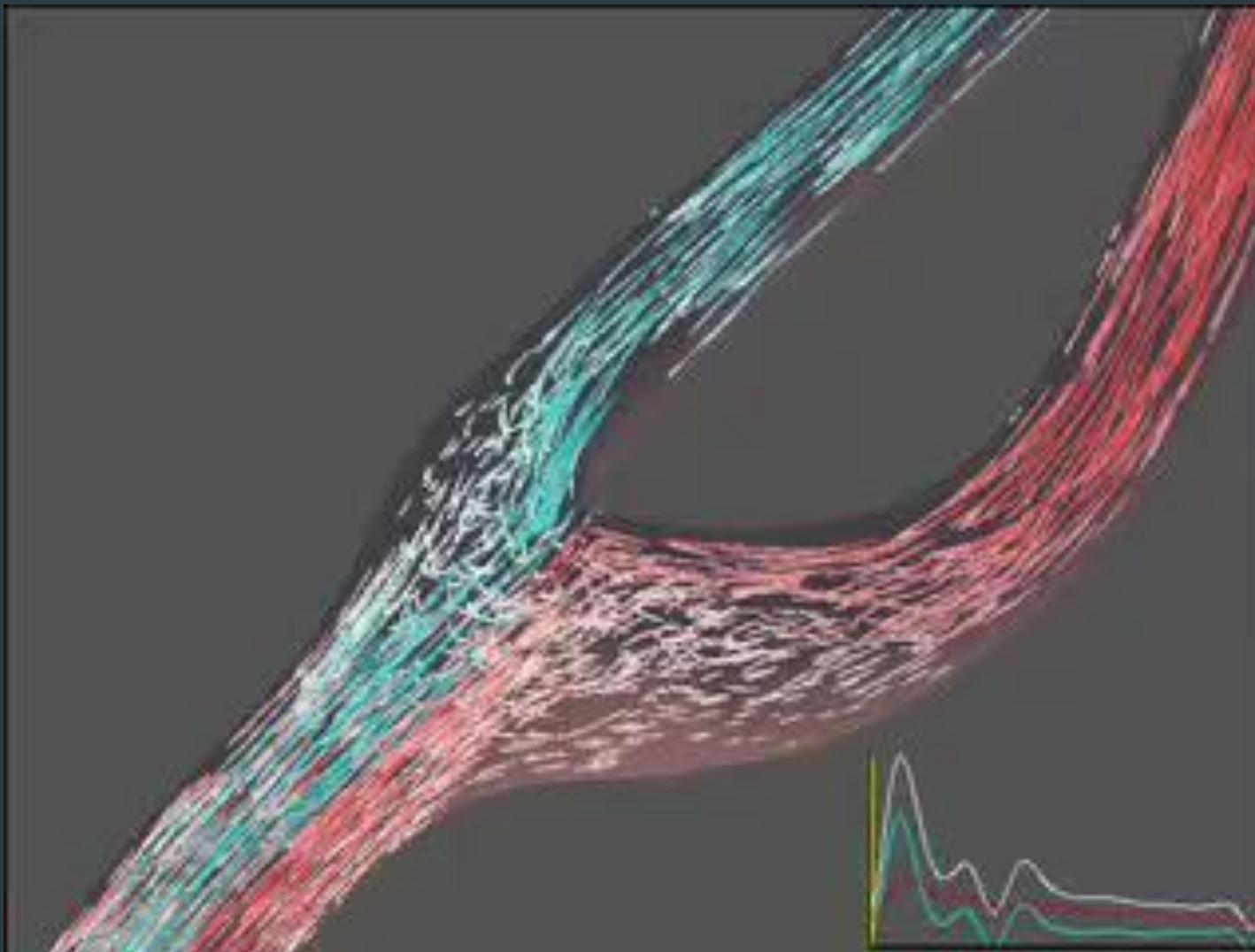
Craig Upson, *Volumetric Visualization Techniques*, in “State of the Art in Computer Graphics: Visualization and Modeling”, David F. Rogers
Rae A. Earnshaw (eds.), Springer-Verlag, 1991, pp. 313-250

Flow visualization of Jupiter



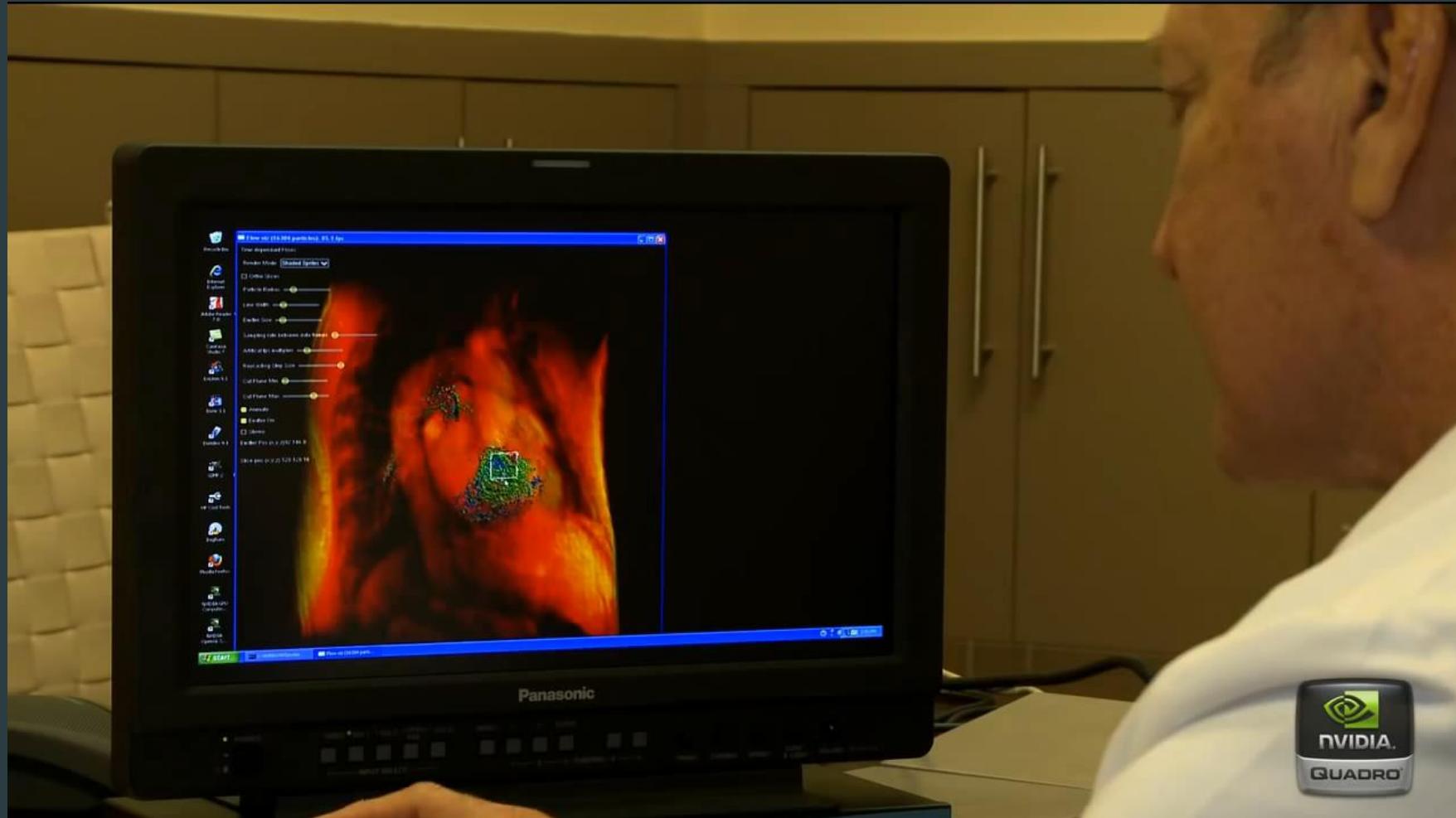
Craig Upson, *Volumetric Visualization Techniques*, in "State of the Art in Computer Graphics: Visualization and Modeling", David F. Rogers Rae A. Earnshaw (eds.), Springer-Verlag, 1991, pp. 313-250

Particle animation



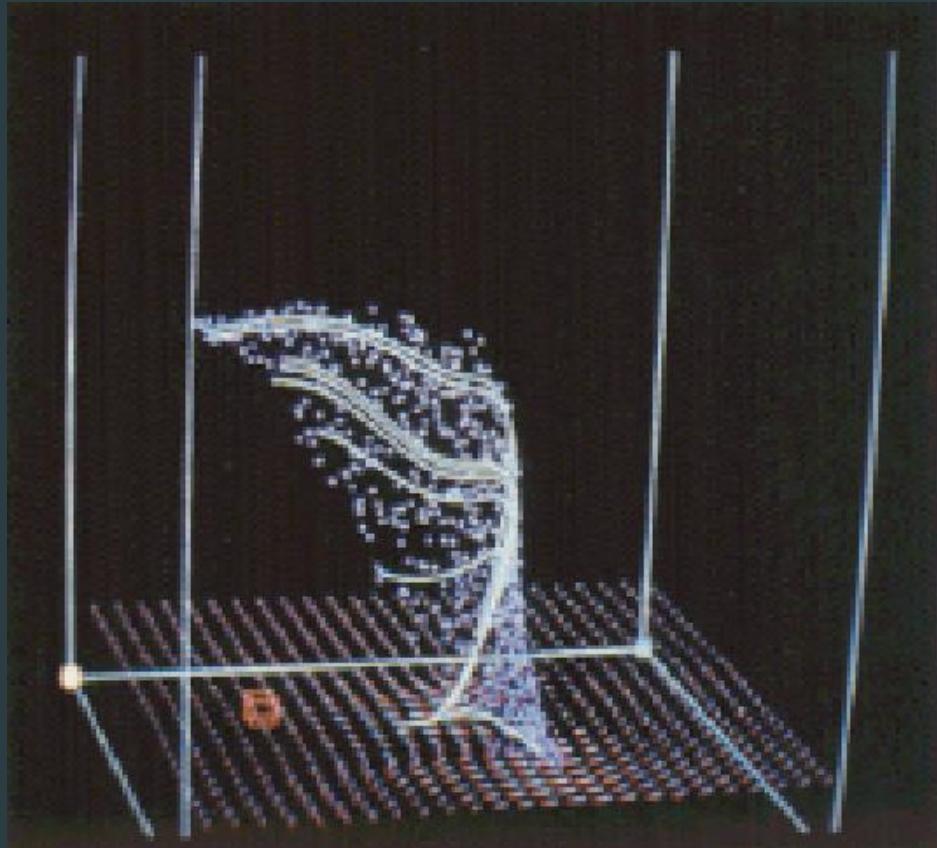
Steinman DA. Simulated pathline visualization of computed periodic blood flow patterns. *J Biomech.* 2000 May;33(5):623-8. doi: 10.1016/s0021-9290(99)00205-5. PMID: 10708784.

Particle animation



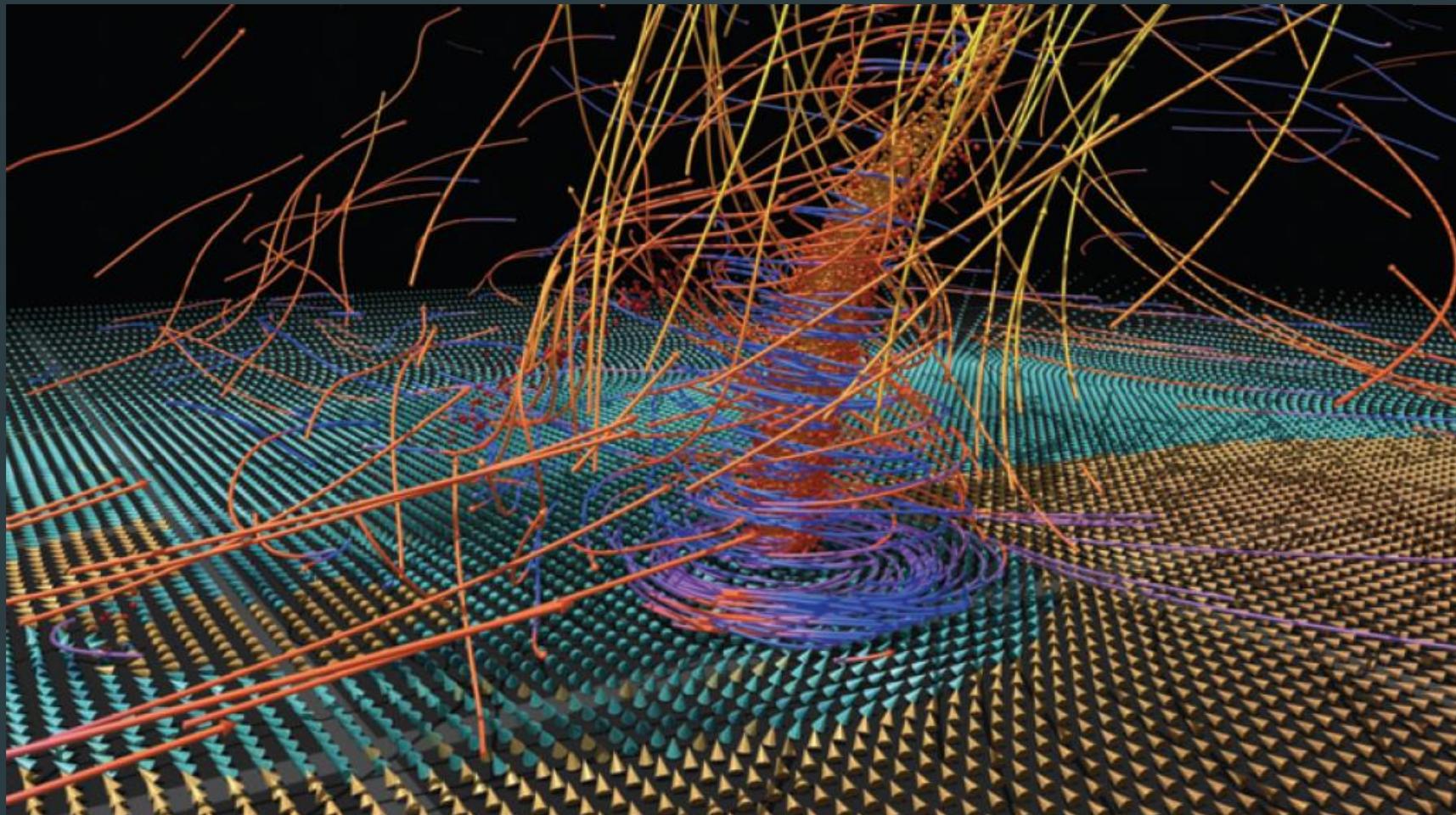
Source: [Computational Visualization of 4D Cardiac Flow](#) (YouTube, 2010)

Particle traces from a numerical simulation of a thunderstorm



Craig Upson, *Volumetric Visualization Techniques*, in "State of the Art in Computer Graphics: Visualization and Modeling", David F. Rogers
Rae A. Earnshaw (eds.), Springer-Verlag, 1991, pp. 313-250

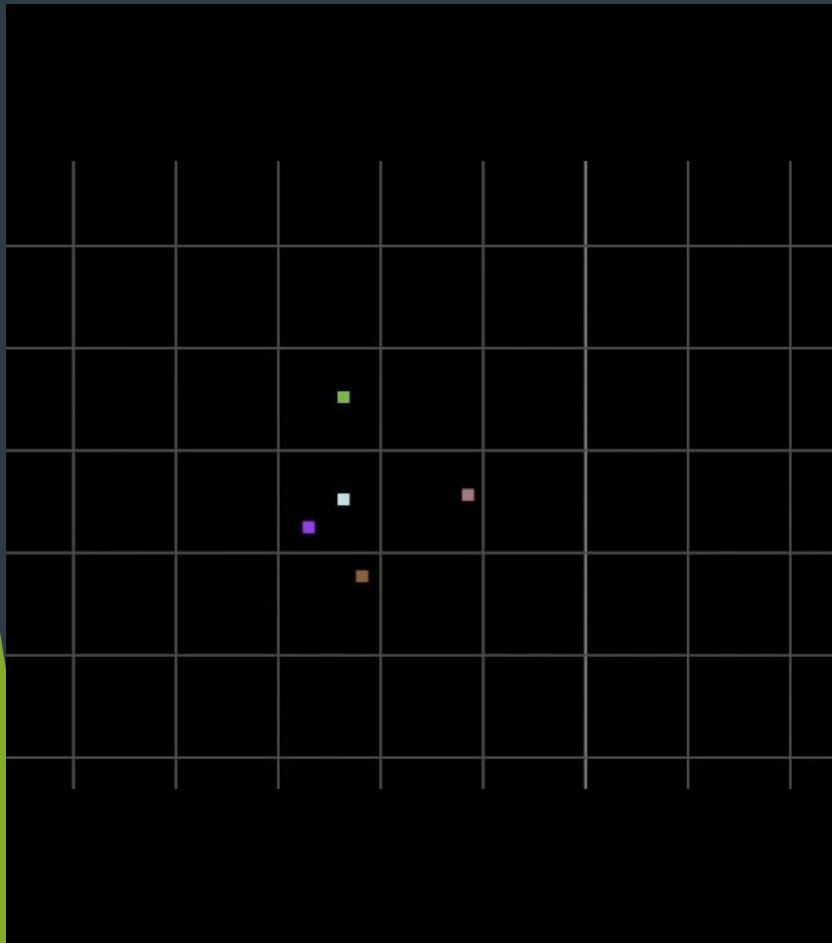
NCSA Visualization of an F3 Tornado



[YouTube link](#). Source: [Advanced Visualization Lab](#), National Center for Supercomputing, University of Illinois

POLDER/NEMO experiment

October 15-28, 2022



POLDER/NEMO experiment

October 15-28, 2022

