

Diffusion Tensor Visualization and the Teem Software that Makes it Go

Gordon Kindlmann
gk@bwh.harvard.edu



Laboratory of Mathematics in Imaging
Department of Radiology
Brigham & Women's Hospital
Harvard Medical School

Outline



Visualization:

DWI Data source

Space of Tensor Shape

Volume Rendering

Superquadric Glyphs

Glyph packing

Outline



Visualization:

DWI Data source

Space of Tensor Shape

Volume Rendering

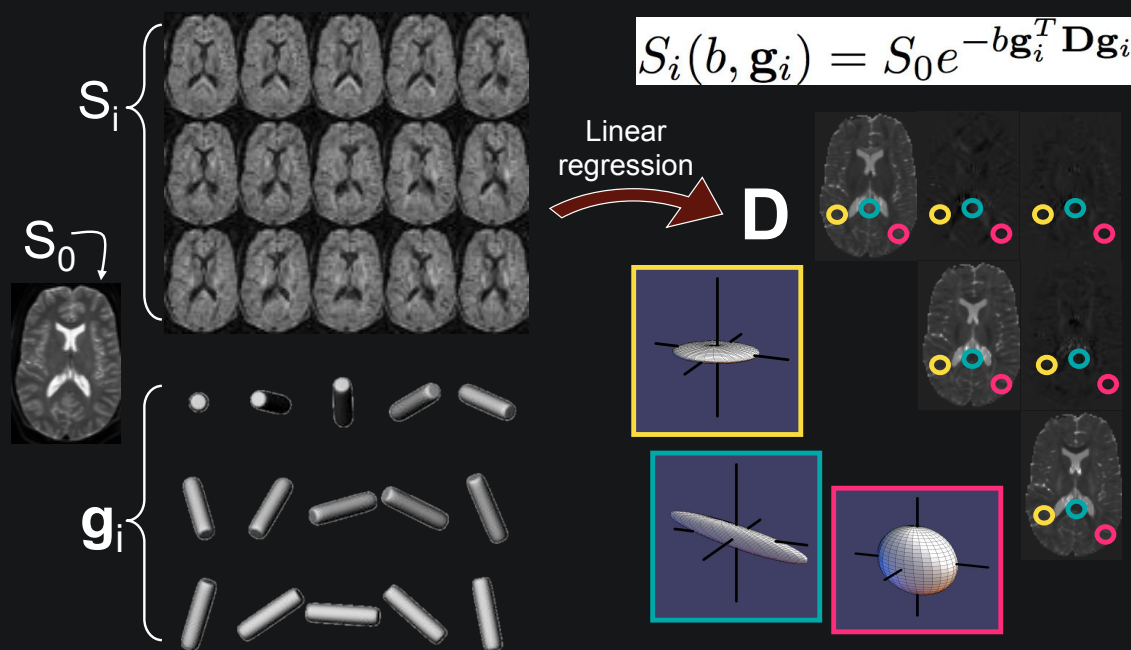
Superquadric Glyphs

Glyph packing

Diffusion weighted, tensor images



Single Tensor Model (Basser et al. 1994)



Outline



Visualization:

DWI Data source

Space of Tensor Shape

Volume Rendering

Superquadric Glyphs

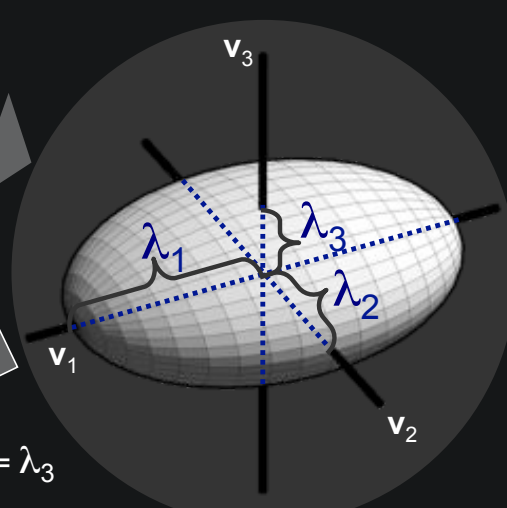
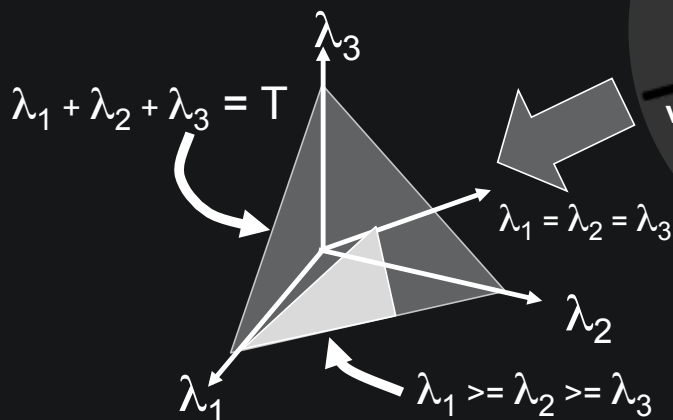
Glyph packing

Eigenvalues == Shape



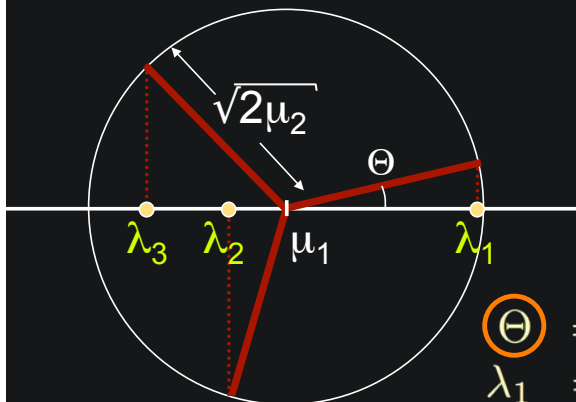
$$D = R \Lambda R^{-1}$$

$$= \begin{bmatrix} | & | & | \\ \mathbf{v}_1 & \mathbf{v}_2 & \mathbf{v}_3 \\ | & | & | \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix} \begin{bmatrix} \text{---} \mathbf{v}_1 \\ \text{---} \mathbf{v}_2 \\ \text{---} \mathbf{v}_3 \end{bmatrix}$$



Tensor shape always has 3 degrees of freedom

Cardano's Formula



Nickalls, 1993

Tensor eigenvalues computed as solutions of a cubic polynomial:

$$\Theta = \cos^{-1}(\sqrt{2}\alpha_3)/3$$

$$\lambda_1 = \mu_1 + \sqrt{2\mu_2}\cos(\Theta)$$

$$\lambda_2 = \mu_1 + \sqrt{2\mu_2}\cos(\Theta - 2\pi/3)$$

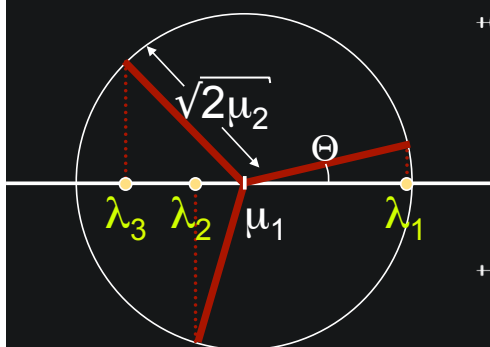
$$\lambda_3 = \mu_1 + \sqrt{2\mu_2}\cos(\Theta + 2\pi/3)$$

μ_1 : mean($\lambda_1, \lambda_2, \lambda_3$)

μ_2 : variance($\lambda_1, \lambda_2, \lambda_3$)

α_3 : skewness($\lambda_1, \lambda_2, \lambda_3$)

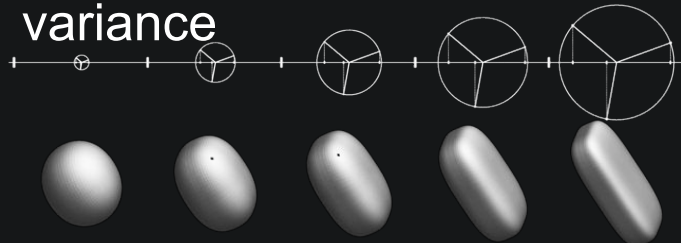
Eigenvalue wheel



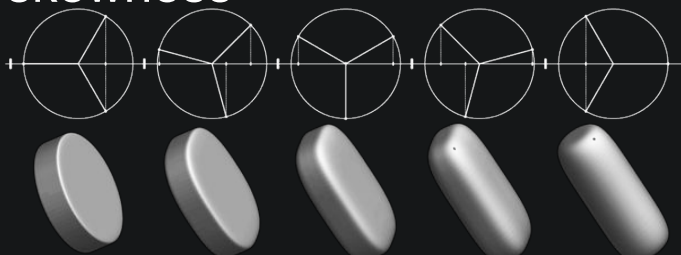
mean



variance



skewness



Eigenvalue "sorting", 2nd order isotropy



$$\Theta = \cos^{-1}(\sqrt{2} \text{ skew})/3 \Rightarrow \Theta \in [0, \pi/3] \Rightarrow \lambda_1 \geq \lambda_2 \geq \lambda_3$$

linear

$$\Theta = 0$$

$$\text{skew} = -1/\sqrt{2}$$

$$\lambda_1 > \lambda_2 = \lambda_3$$

"orthotropic"

$$\Theta = \pi/6$$

$$\text{skew} = 0$$

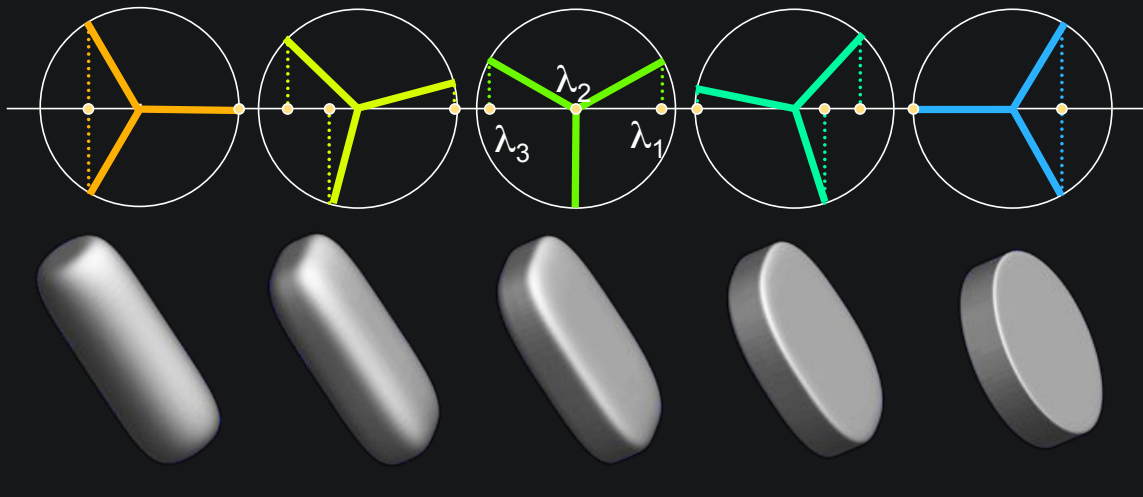
$$\lambda_1 - \lambda_2 = \lambda_2 - \lambda_3$$

planar

$$\Theta = \pi/3$$

$$\text{skew} = 1/\sqrt{2}$$

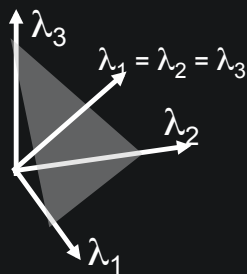
$$\lambda_1 = \lambda_2 > \lambda_3$$



Tensor invariants as orthogonal shape parameterizations



Cylindrical or spherical coordinates (Ennis & Kindlmann 2005)

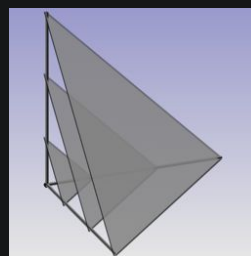


$$\text{tr}(\mathbf{D}) = D_{xx} + D_{yy} + D_{zz}$$

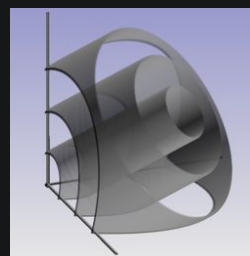
$$|\mathbf{D}| = \sqrt{\text{tr}(\mathbf{D}^T \mathbf{D})}$$

$$\mathbf{E} = \text{deviatoric}(\mathbf{D})$$

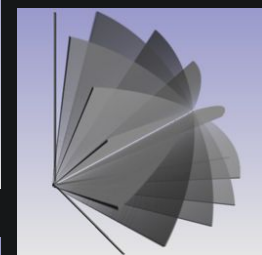
$$= \mathbf{D} - \text{trace}(\mathbf{D}) \cdot \mathbf{I}/3$$



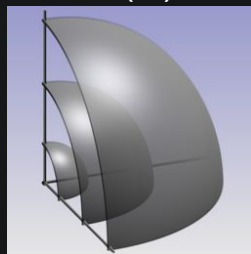
$\text{tr}(\mathbf{D})$



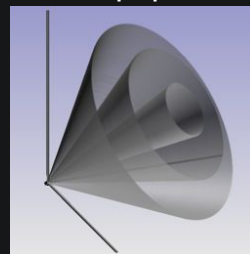
$|\mathbf{D}|$



$\text{mode}(\mathbf{E})$
 $= \det(\mathbf{E}/|\mathbf{E}|)$
 (Criscione '00)
 Mode measures
 Linear vs. planar
 anisotropy



$|\mathbf{D}|$

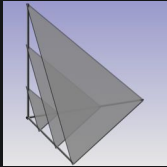


$|\mathbf{E}|/|\mathbf{D}| \sim \text{FA}$
 FA = Fractional
 Anisotropy

Biological Meaning of Tensor Shape

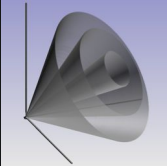


Size: **bulk mean diffusivity** (“ADC”)



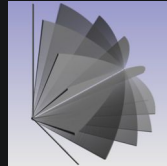
- ADC strictly speaking diffusivity along **one** direction
- Note: same across gray+white matter, high in CSF
- Indicator of acute ischemic stroke

• Anisotropy (e.g. FA): directional microstructure



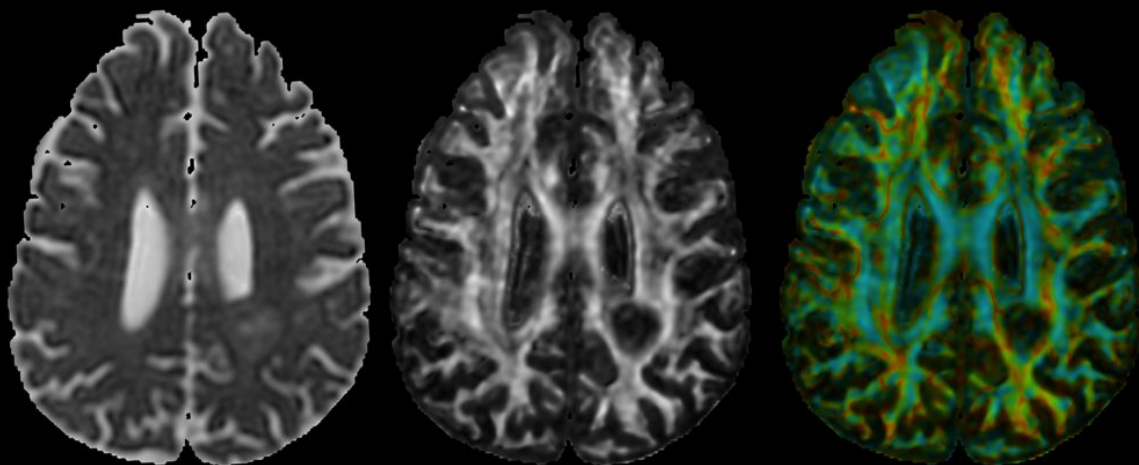
- High in white matter, low in gray matter and CSF
- Increases with myelination, decreases in some diseases (Multiple Sclerosis)

• Mode: linear versus planar



- Partial voluming of adjacent orthogonal structures
- Fine-scale mixing of diverse fiber directions
- Tensor fitting error increases with planarity (Tuch 2002)

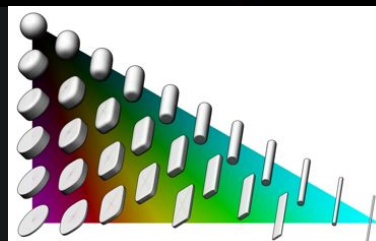
Tensor shape on one slice

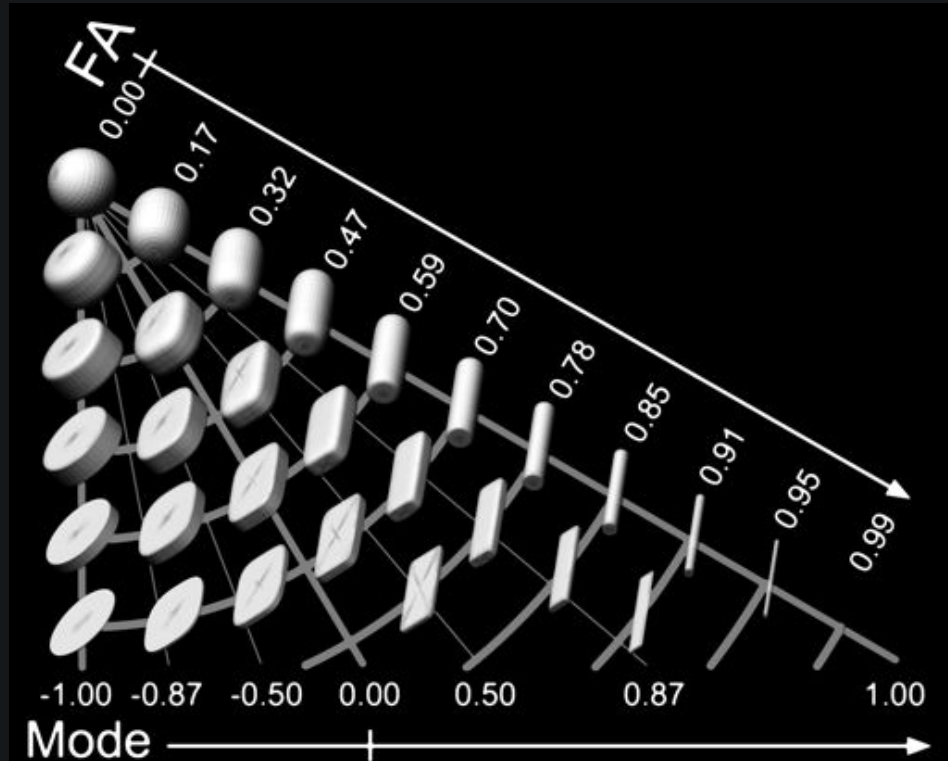


Trace

Fractional Anisotropy

“Anisotropy” is a bivariate quantity





Visualization:

DWI Data source

Space of Tensor Shape

Volume Rendering

Superquadric Glyphs

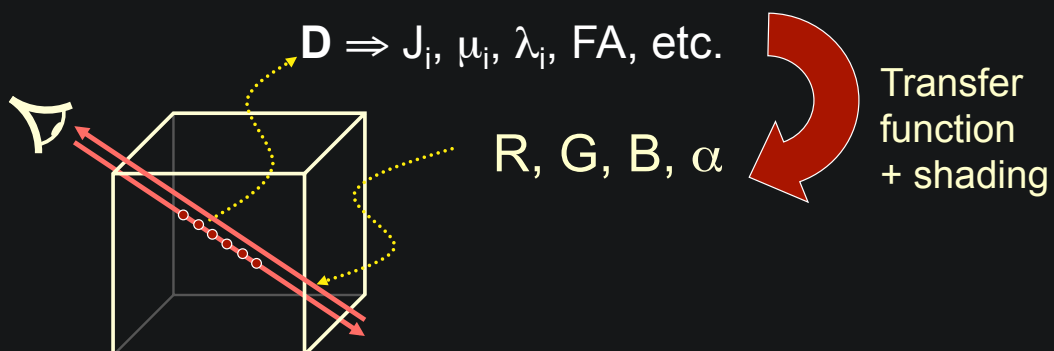
Glyph packing

Direct Volume Rendering

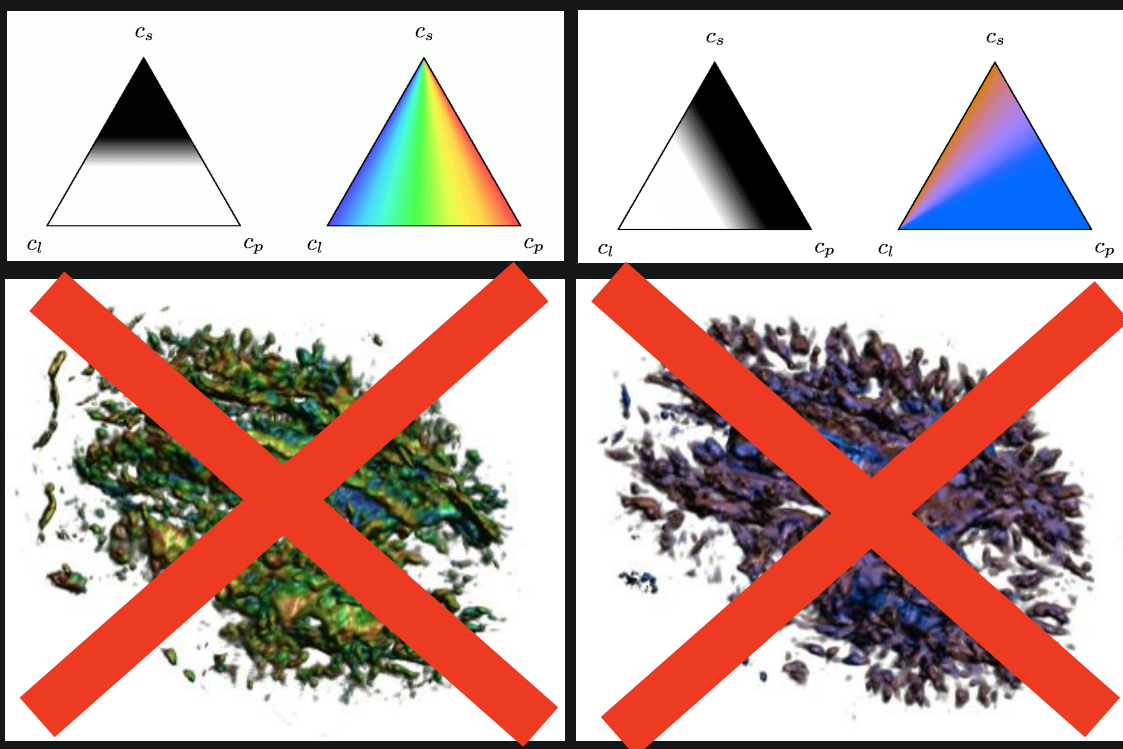


Simple algorithm

- Cast rays through volume
- Measure tensor, tensor properties
- Assign colors and opacities
- Modulate colors with shading
- Composite with “over” operator



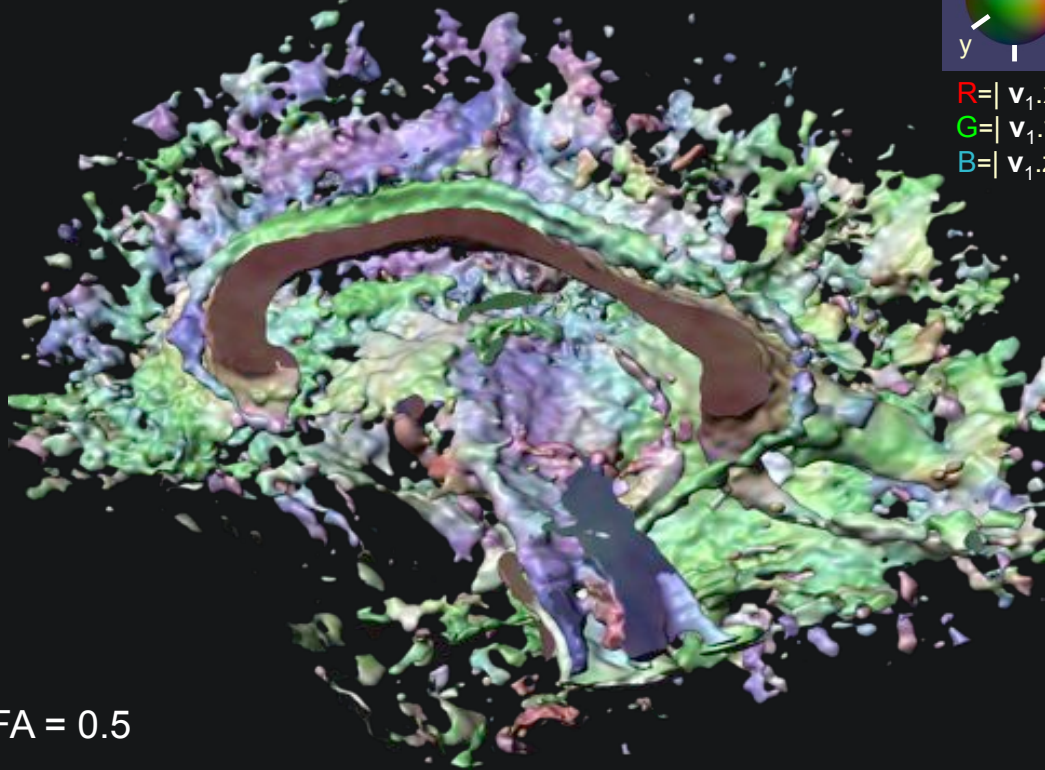
Barycentric color maps: results



Volume Rendering Results

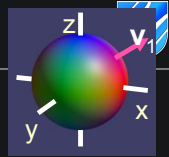


$$R = |v_1 \cdot x|$$
$$G = |v_1 \cdot y|$$
$$B = |v_1 \cdot z|$$

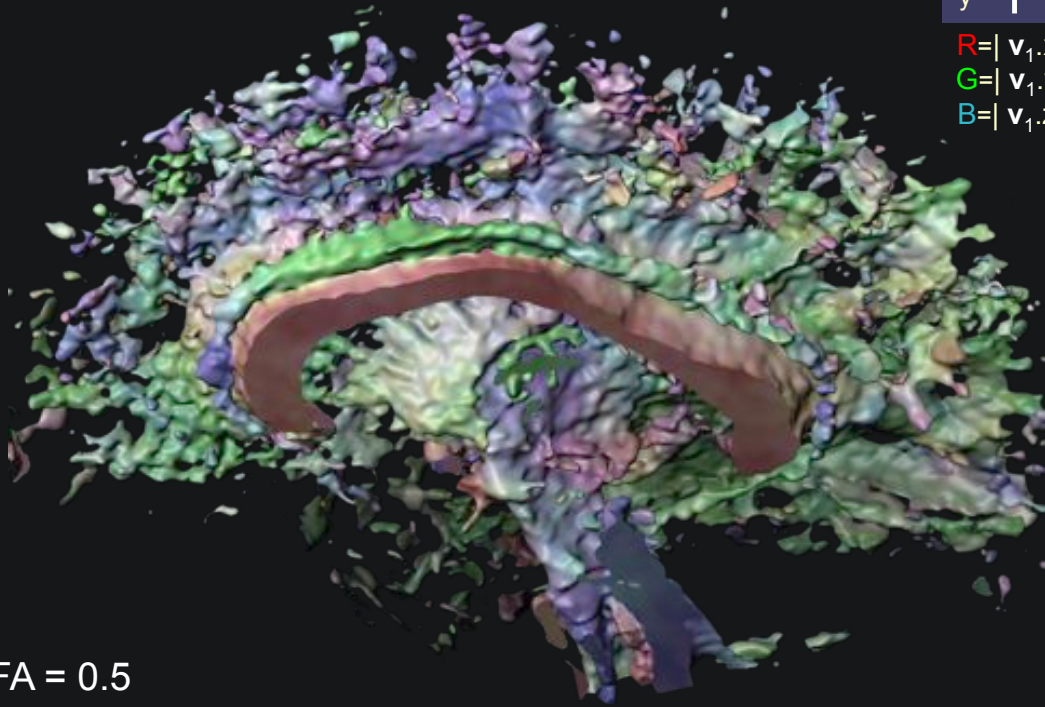


FA = 0.5

Volume Rendering Results



$$R = |v_1 \cdot x|$$
$$G = |v_1 \cdot y|$$
$$B = |v_1 \cdot z|$$



FA = 0.5

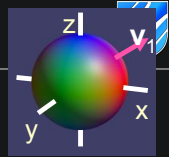
Volume Rendering Results



$$R = |v_1 \cdot x|$$
$$G = |v_1 \cdot y|$$
$$B = |v_1 \cdot z|$$

FA = 0.5

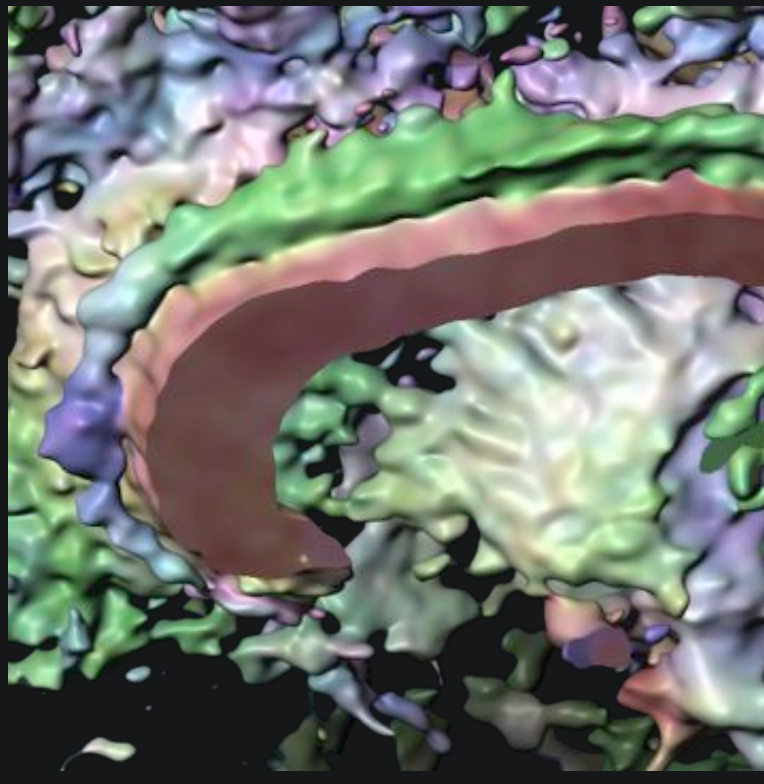
Volume Rendering Results



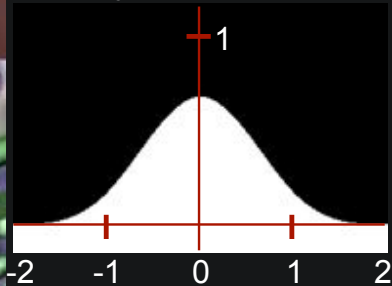
$$R = |v_1 \cdot x|$$
$$G = |v_1 \cdot y|$$
$$B = |v_1 \cdot z|$$

FA = 0.5

Visualizing kernel differences

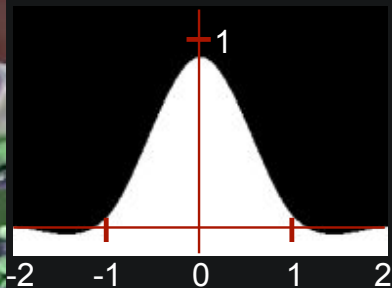
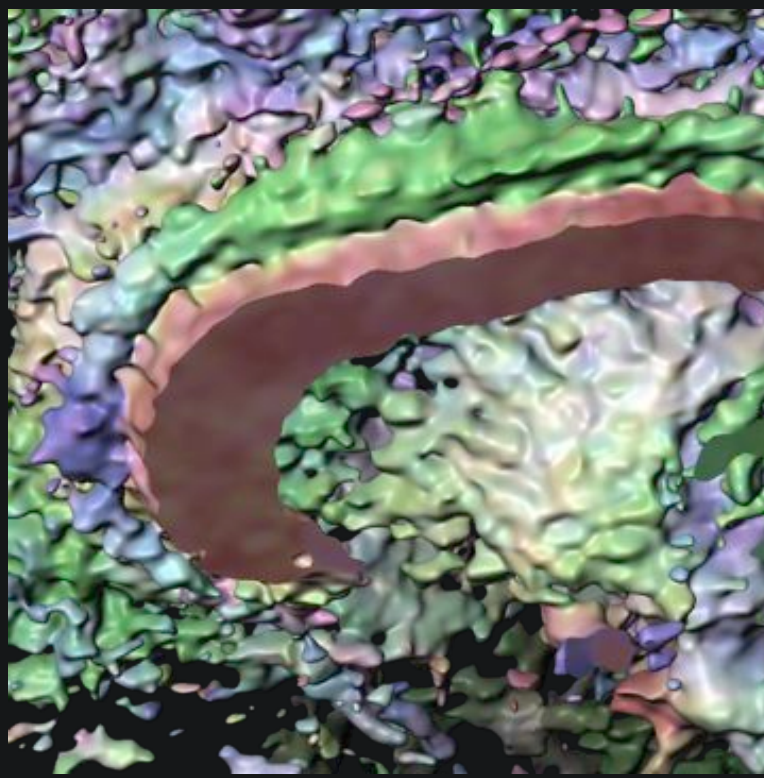


Mitchell-Netravali
BC-splines:
Simple, tunable,
always C^1



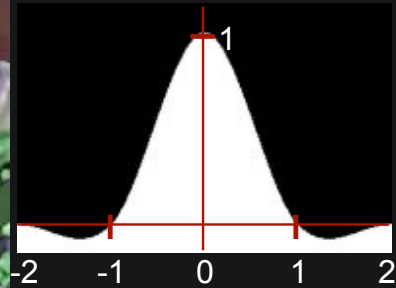
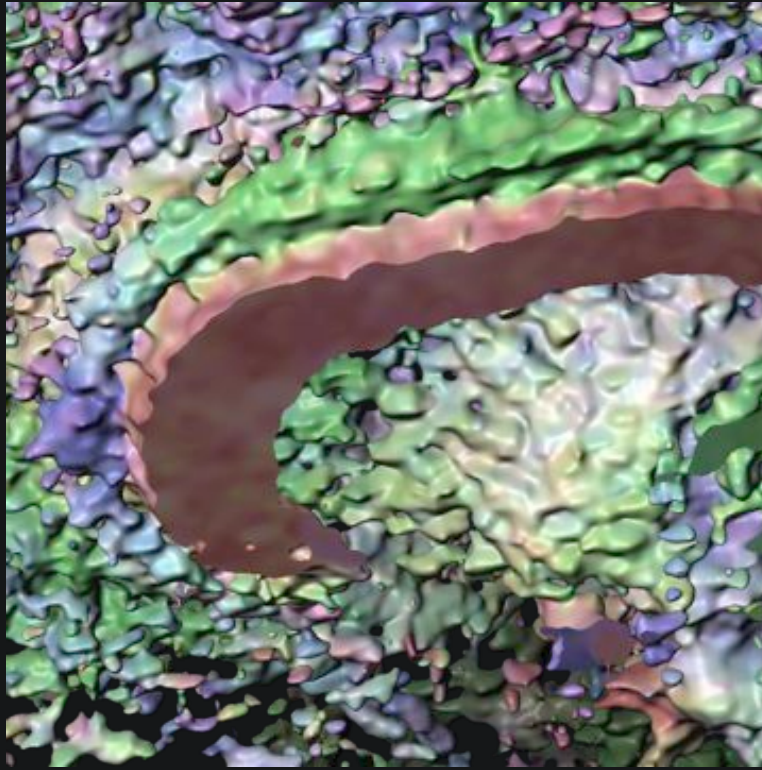
$(B,C) = (1,0)$
Uniform cubic
B-spline, also C^2

Visualizing kernel differences



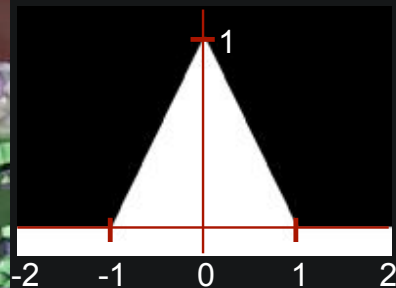
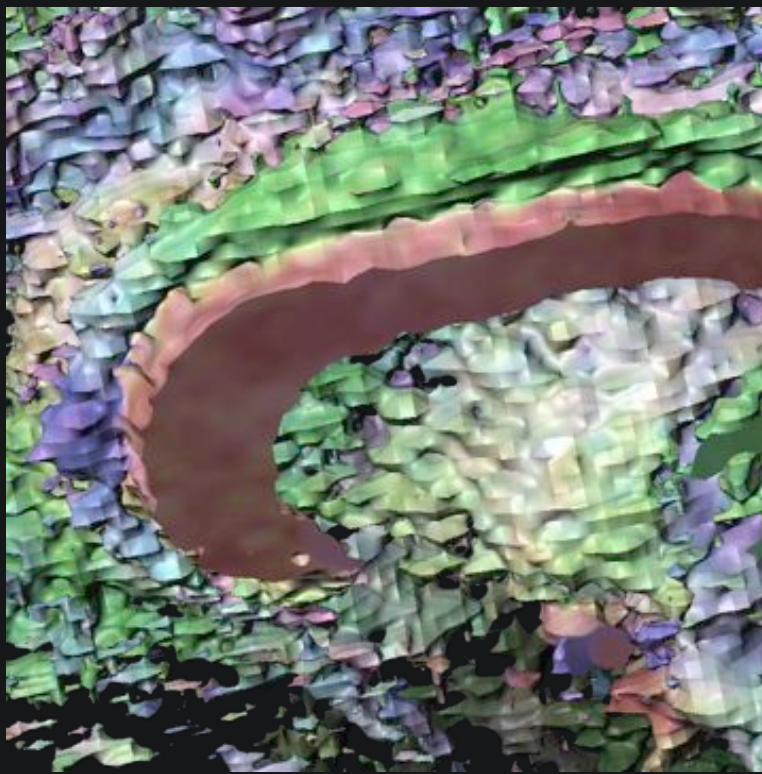
$(B,C) = (1/3, 1/3)$
Blurs a little

Visualizing kernel differences



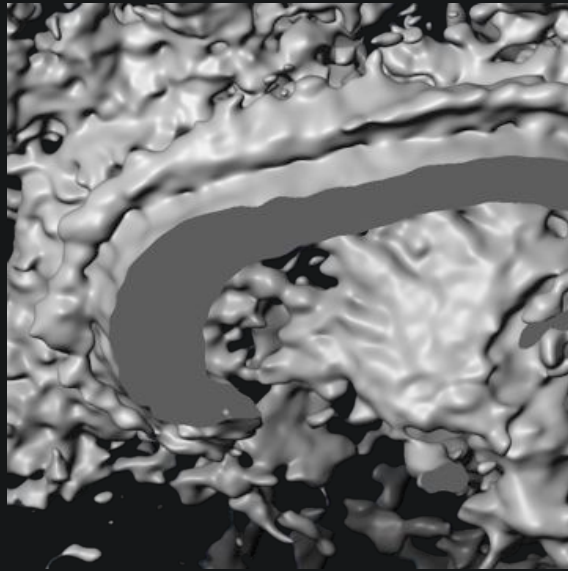
$(B,C) = (0, 1/2)$
Catmull-Rom
Interpolates

Visualizing kernel differences

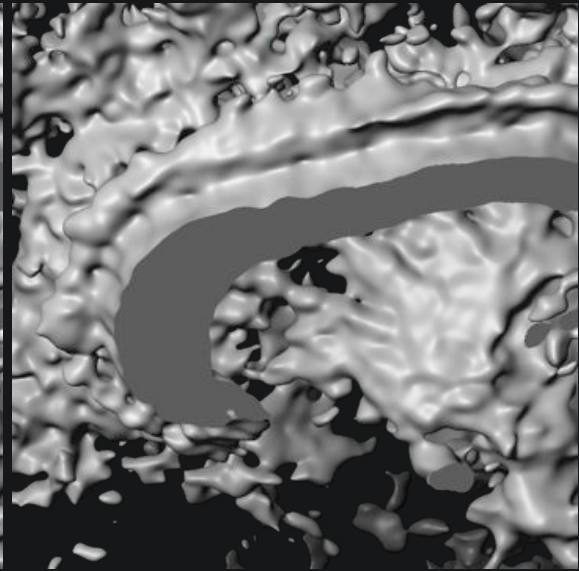


Linear : Not C^1
 \Rightarrow nasty edges,
but can see
each sample

Reconstruction+invariants don't commute

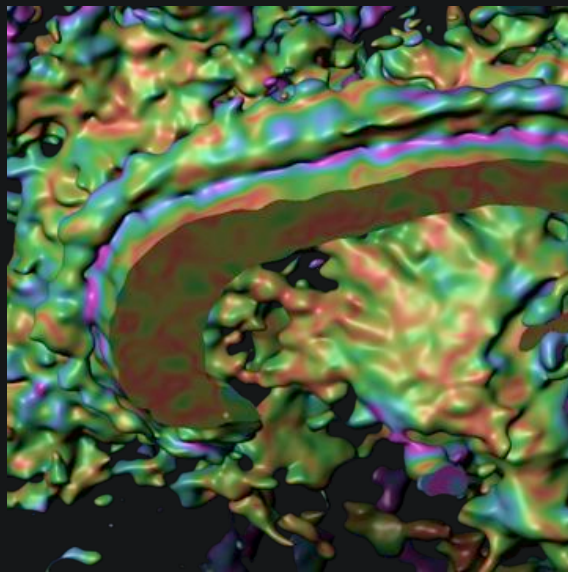


Reconstruct tensors, then
Calculate FA

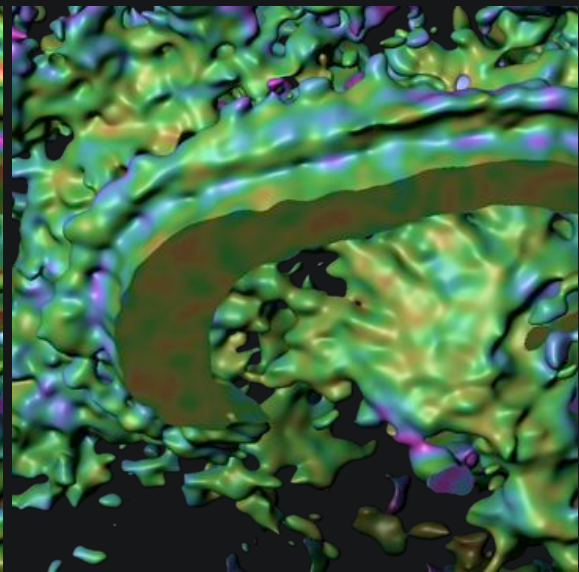


Calculate FA, then
Reconstruct FAs

Reconstruction+invariants don't commute



Reconstruct tensors, then
Calculate FA and Skew



Calculate FA and Skew, then
Reconstruct FAs and Skews



Outline



Visualization:

DWI Data source

Space of Tensor Shape

Volume Rendering

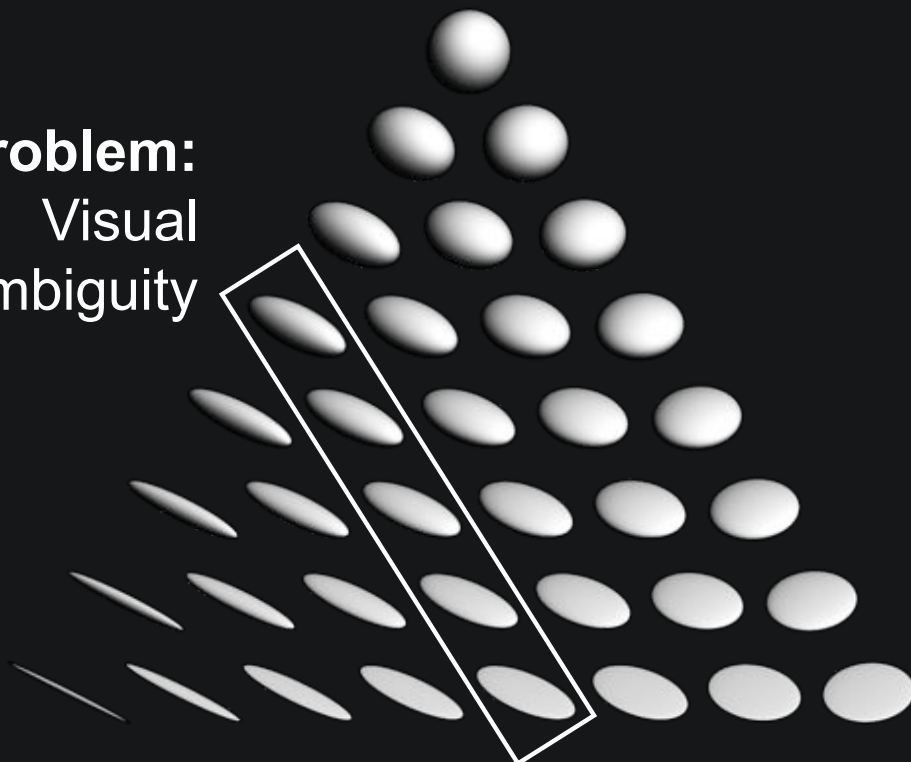
Superquadric Glyphs

Glyph packing

Glyphs: ellipsoids



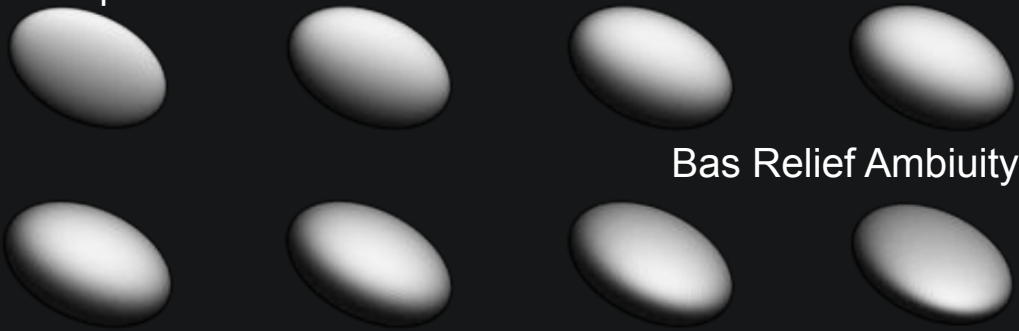
Problem:
Visual
ambiguity



Worst case scenario: ellipsoids

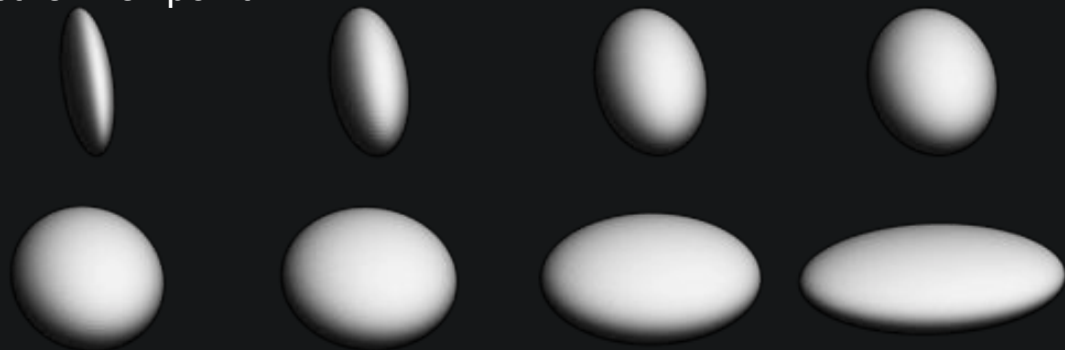


one viewpoint:



Bas Relief Ambiguity

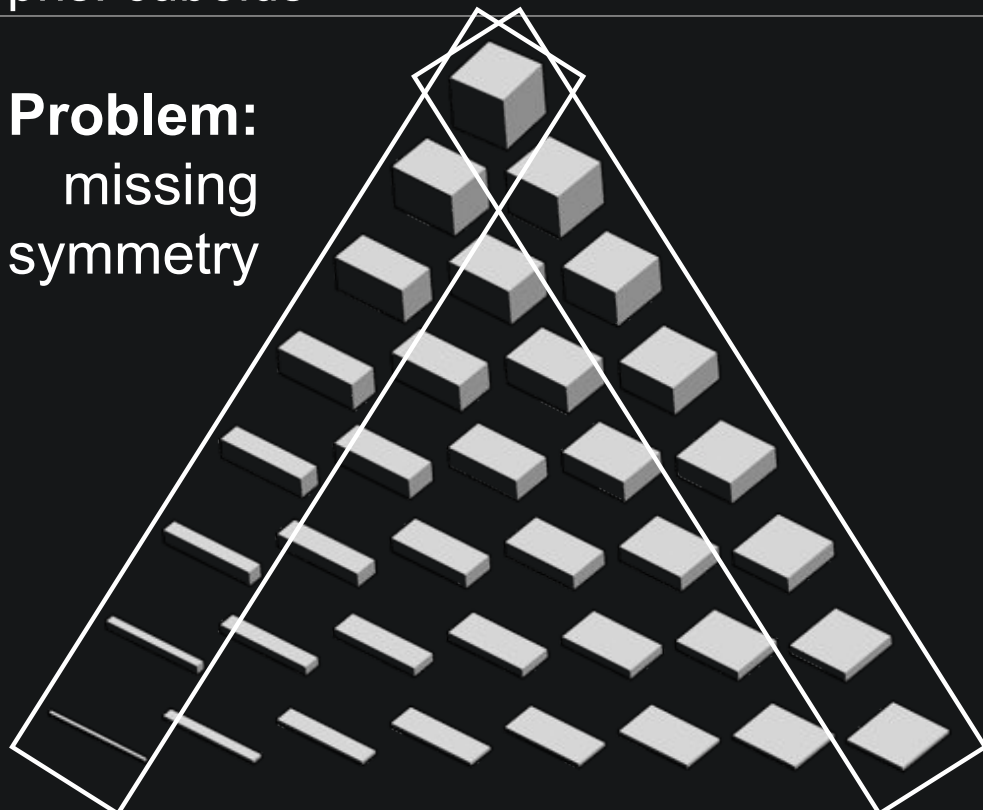
another viewpoint:



Glyphs: cuboids



Problem:
missing
symmetry



Eigensystem symmetry



$$\lambda_1 = \lambda_2$$



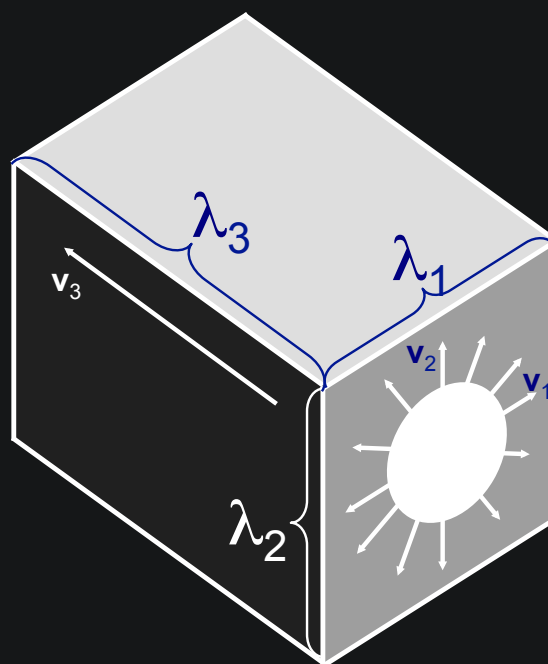
2-D eigenspace(λ_1)



$\mathbf{v}_1, \mathbf{v}_2$ not unique
(continuous rotational symmetry)

Conversely,

$\mathbf{v}_1, \mathbf{v}_3$ distinct

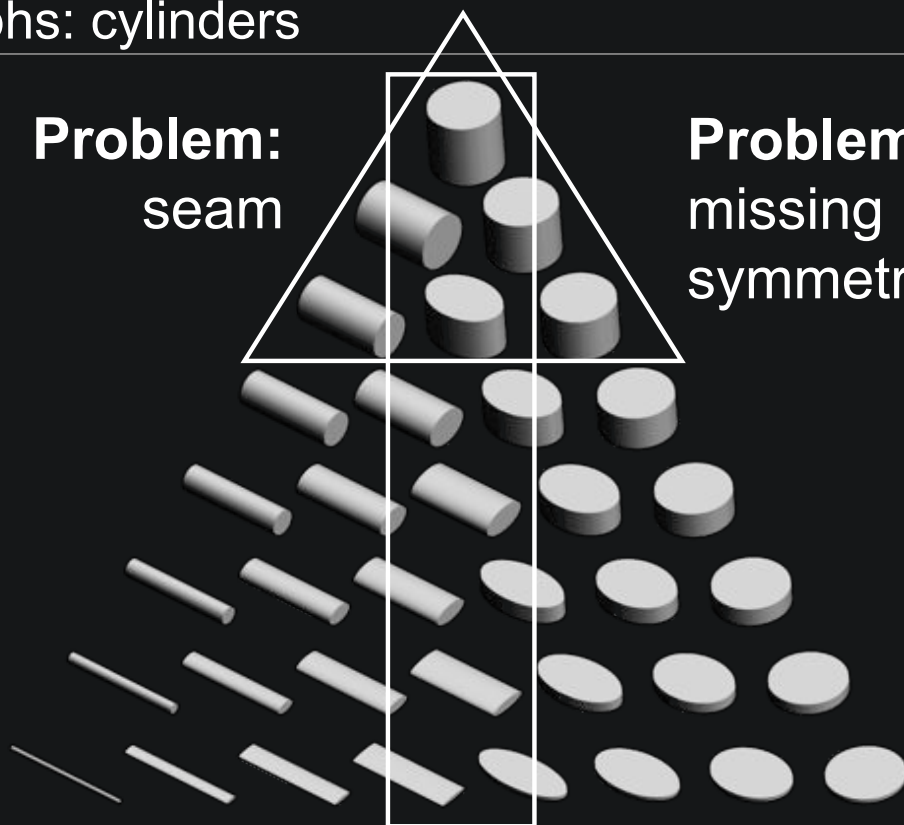


Glyphs: cylinders



Problem:
seam

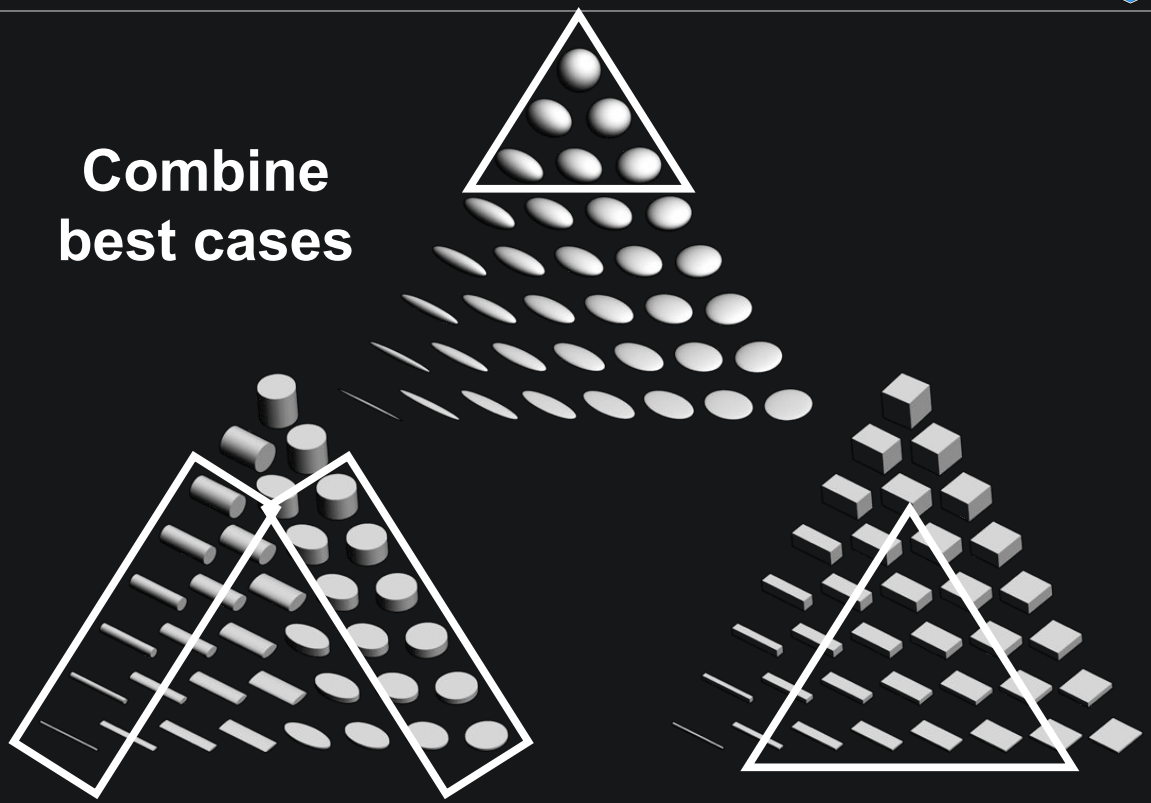
Problem:
missing
symmetry



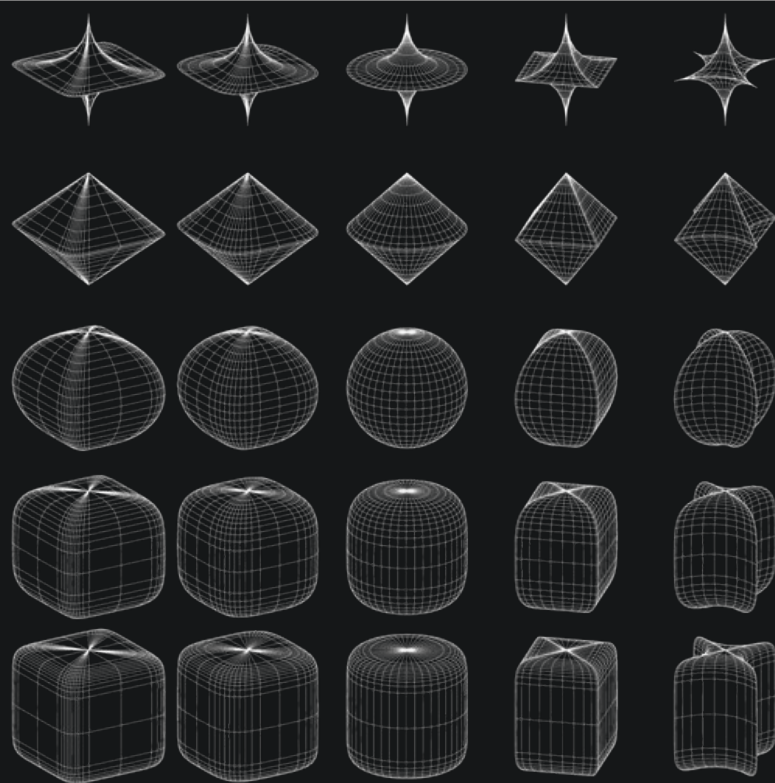
Basic idea:



Combine
best cases



How: superquadrics



Barr, 1981

For visualization:
Shaw et al.,
1998, 1999

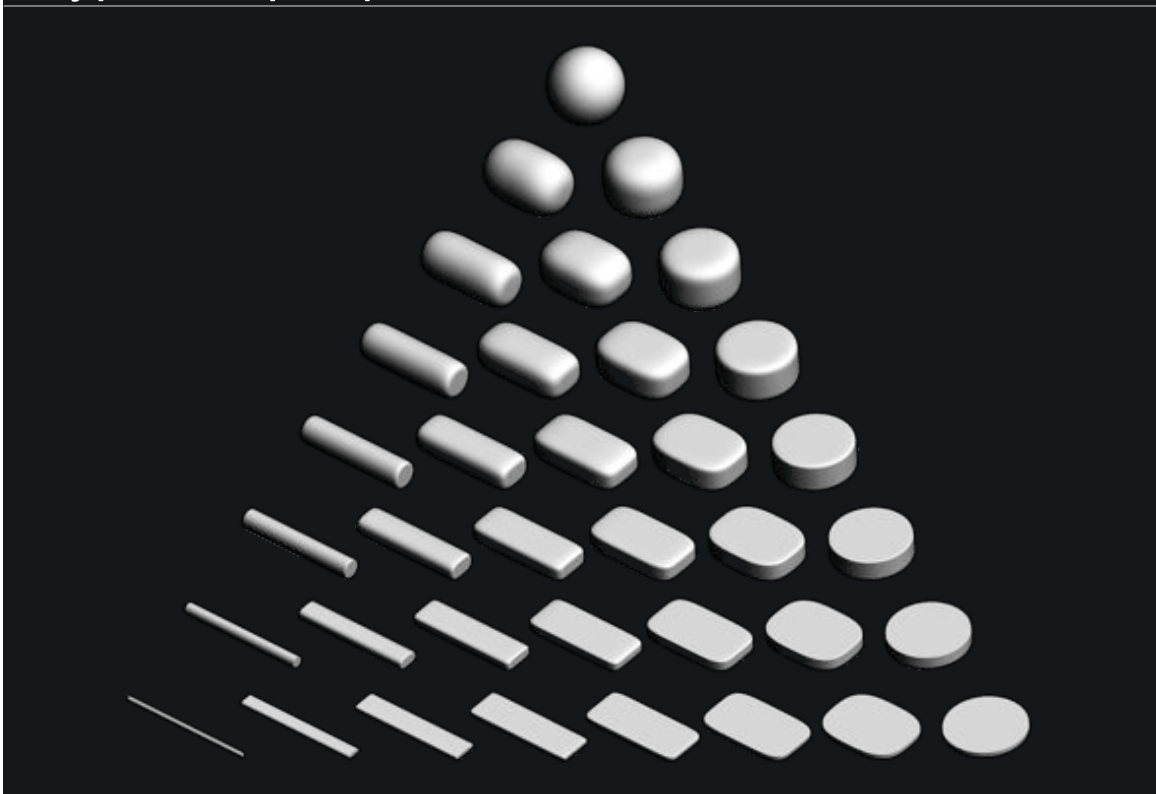
Ebert et al.,
2000, 2001

- Variable glyph geometry
- Tensor glyph, not just multi-value

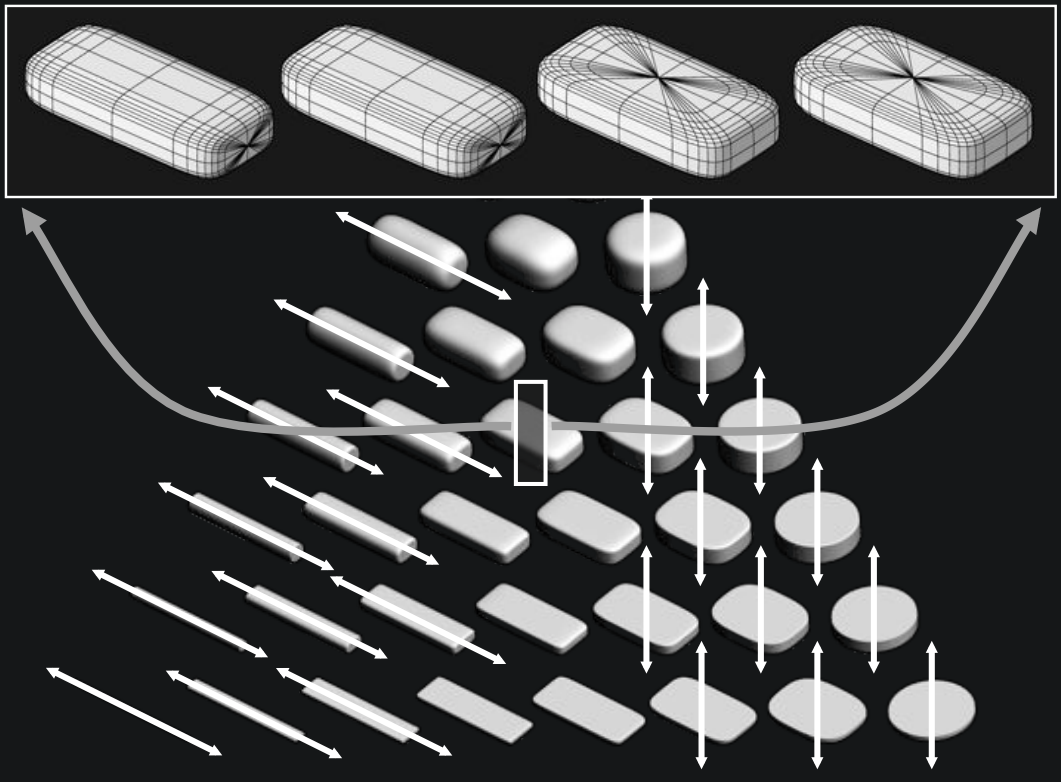
Basic method



Glyphs: superquadrics



Seam: hidden

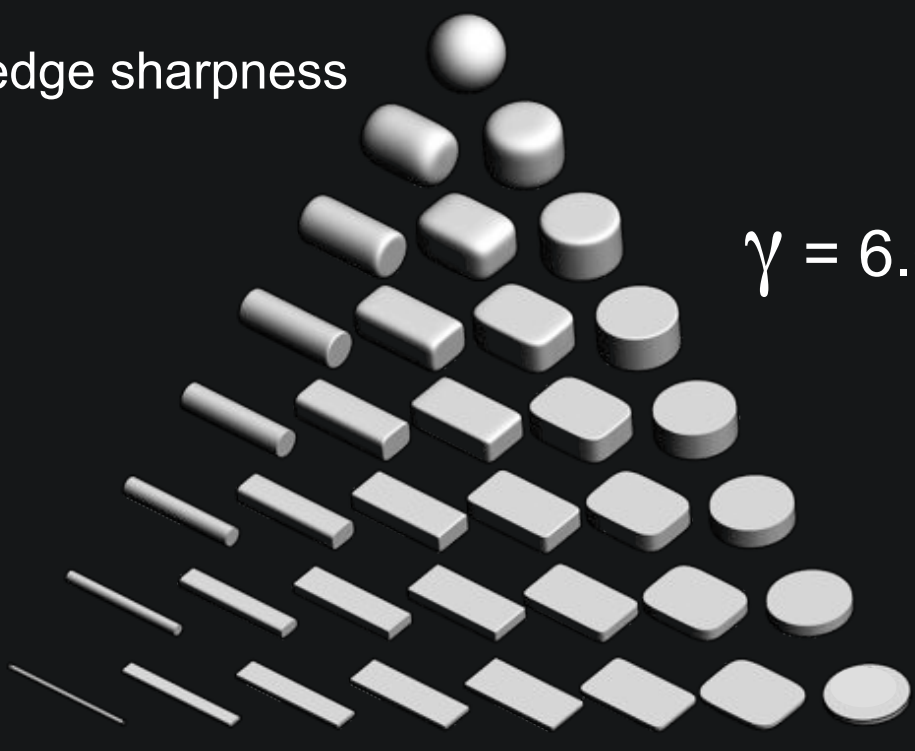


One parameter

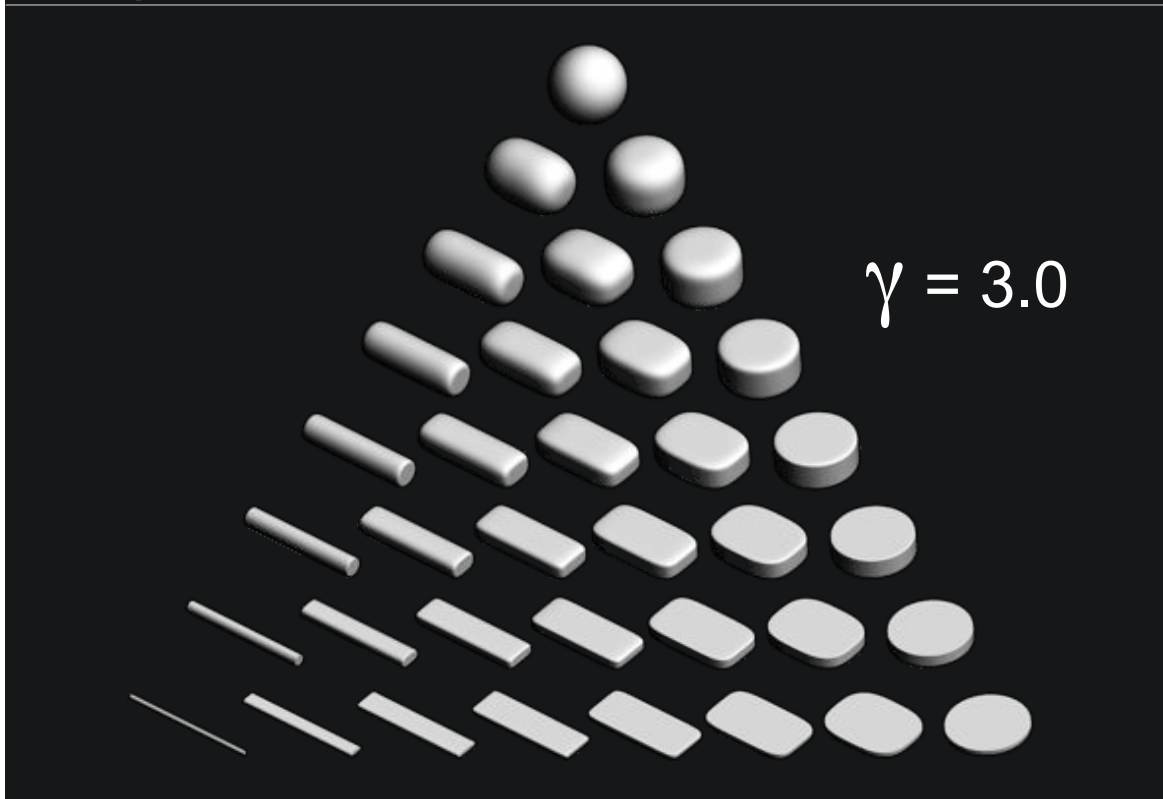


γ : edge sharpness

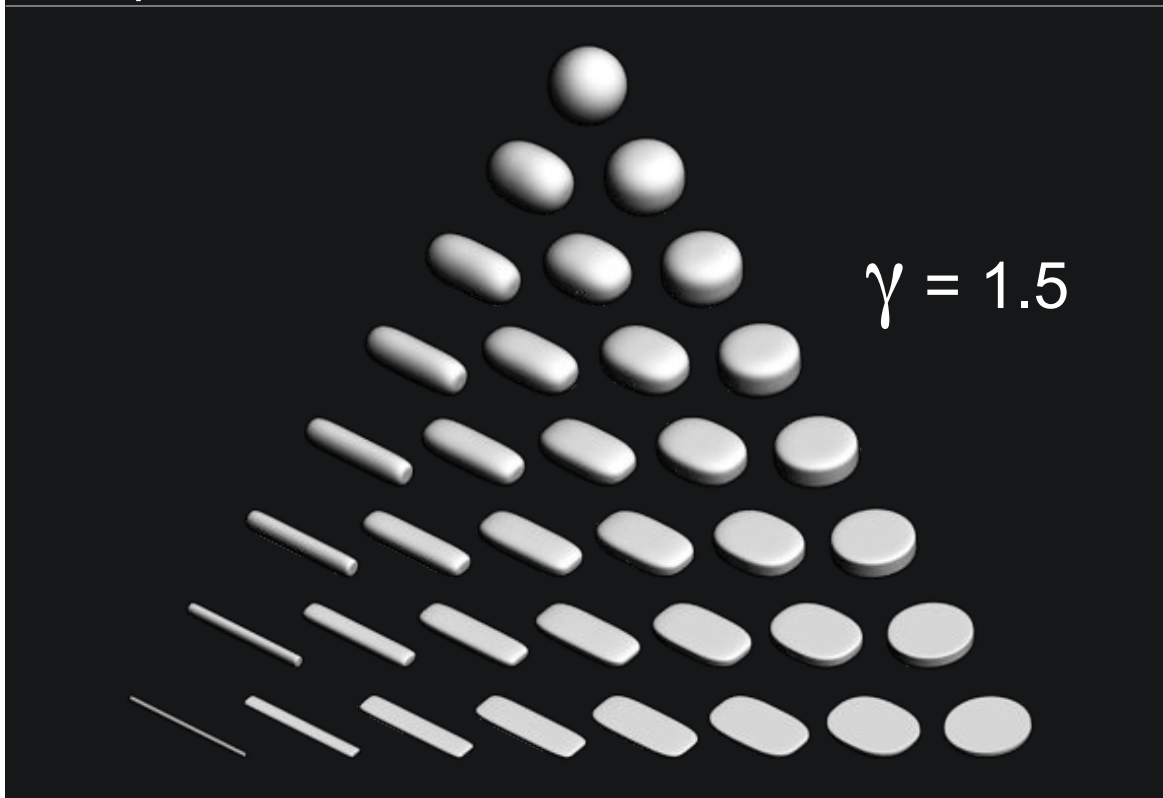
$$\gamma = 6.0$$



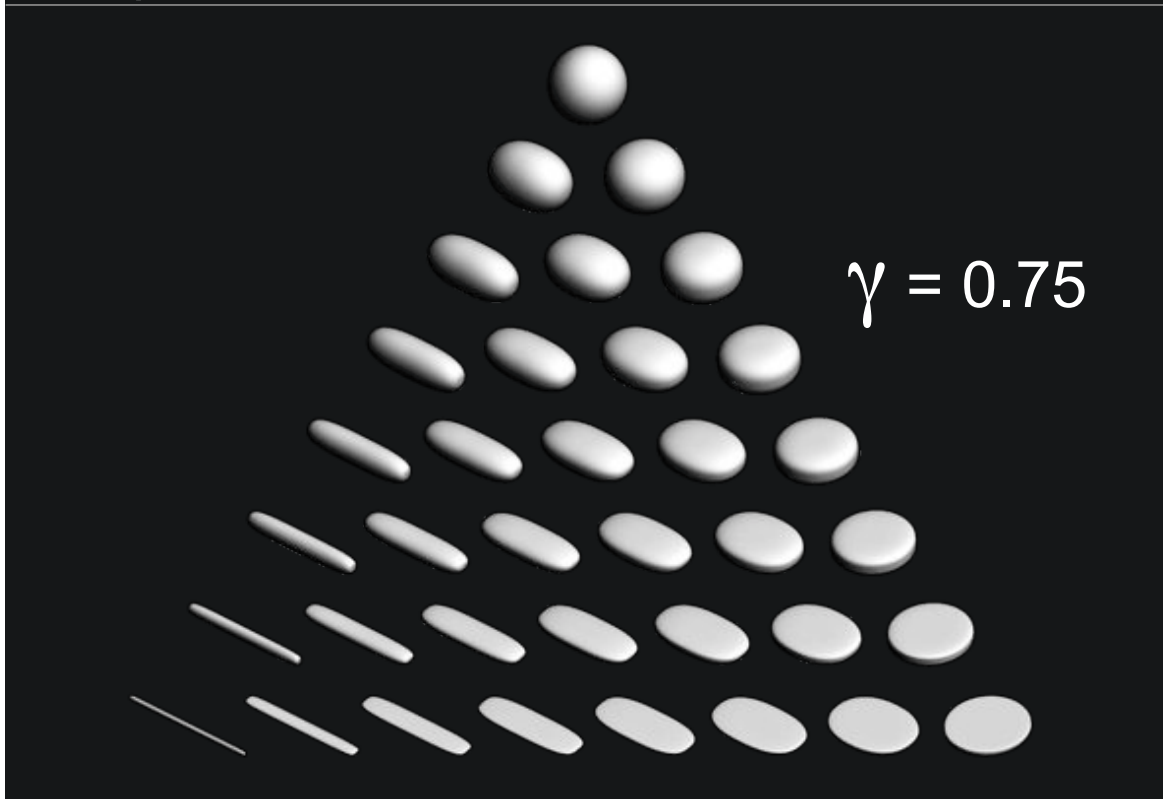
One parameter



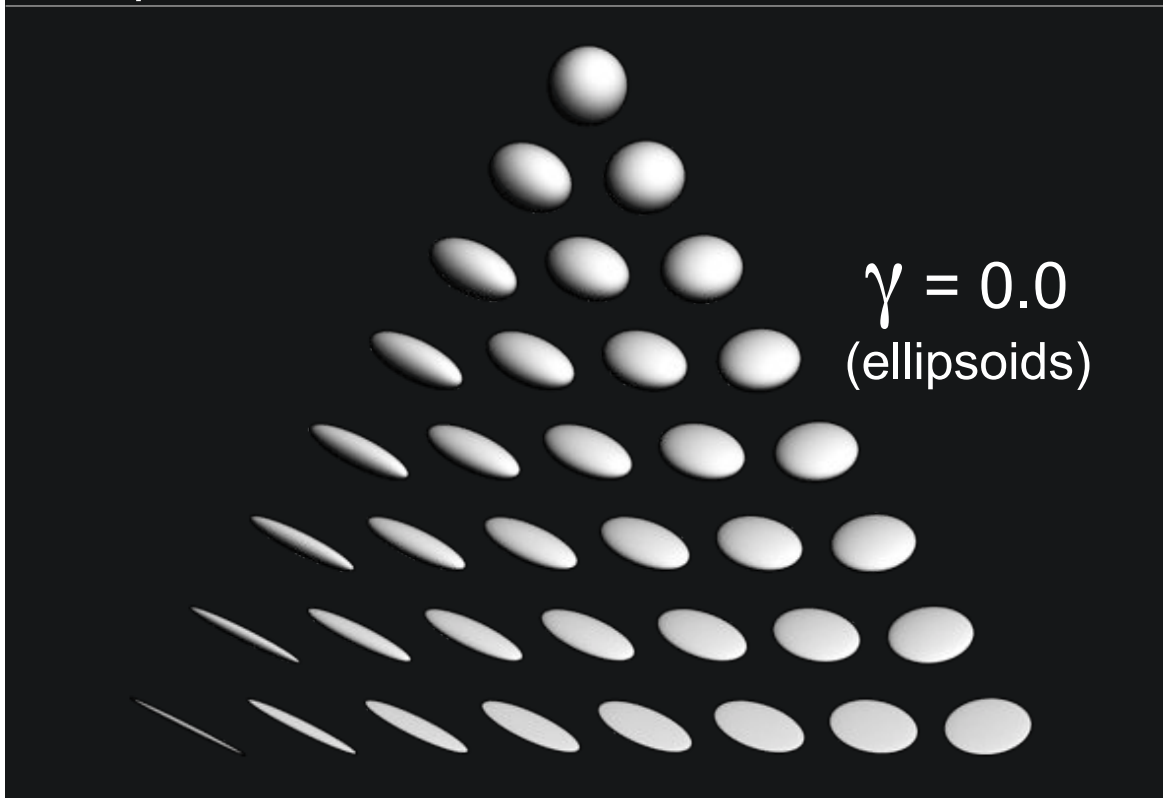
One parameter



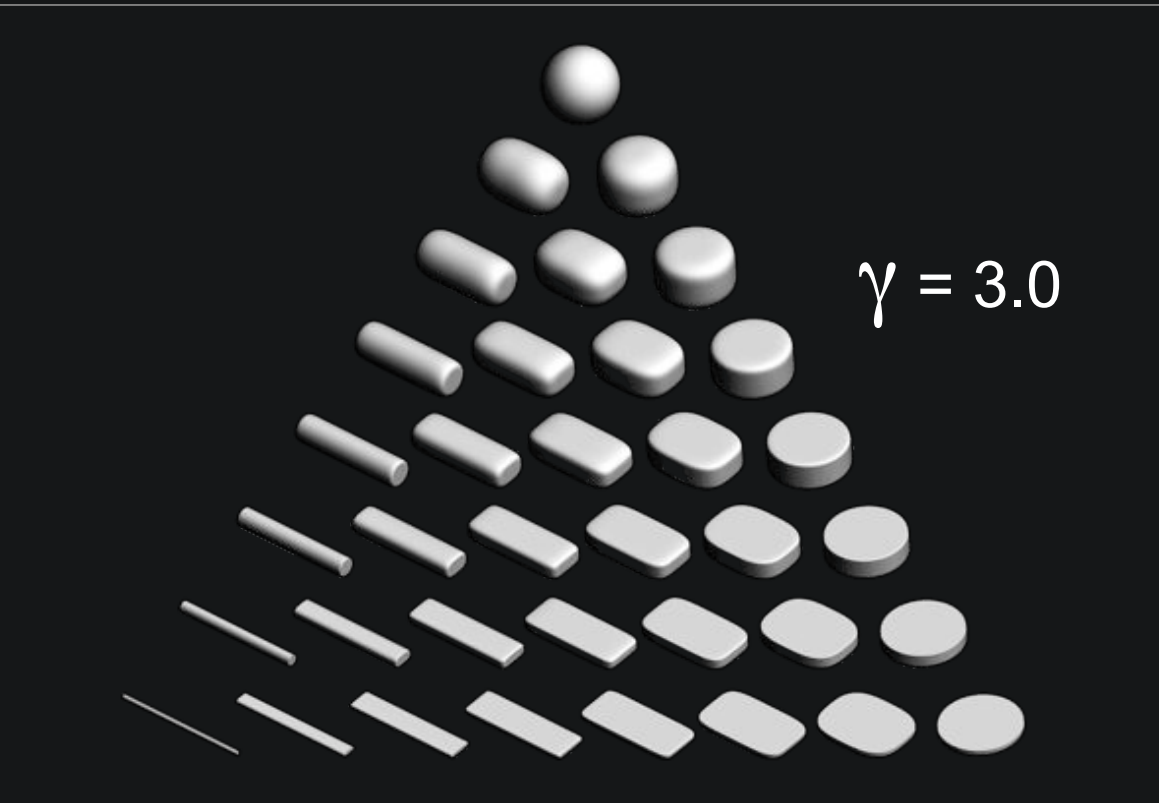
One parameter



One parameter



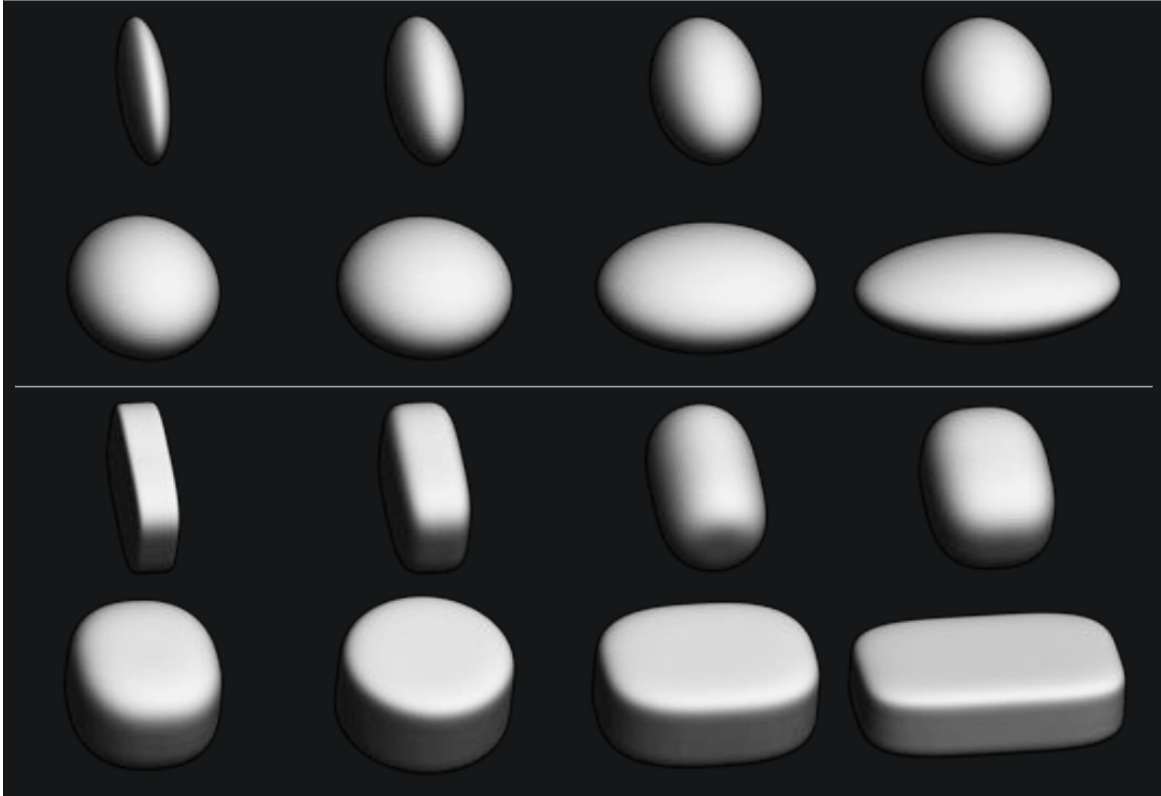
Good default choice



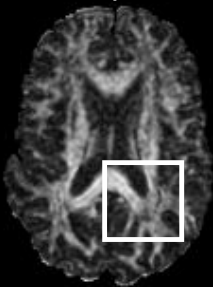
Worst case scenario, revisited



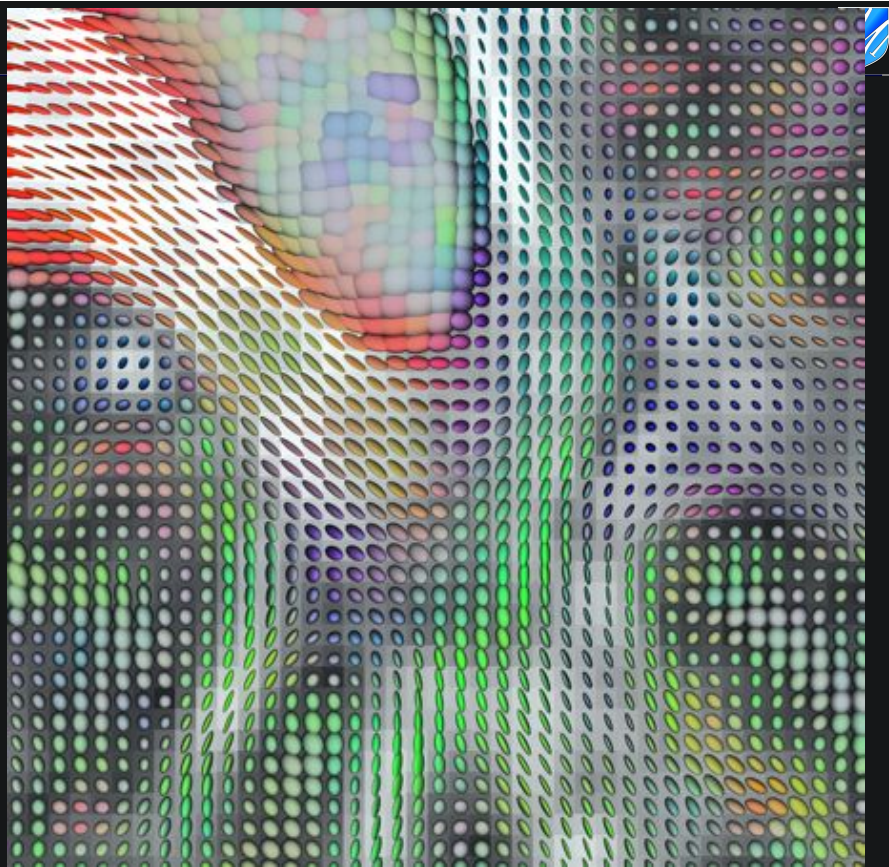
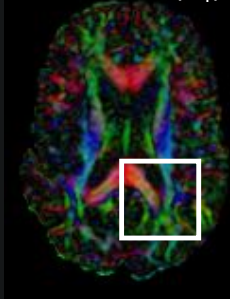
Worst case scenario, revisited

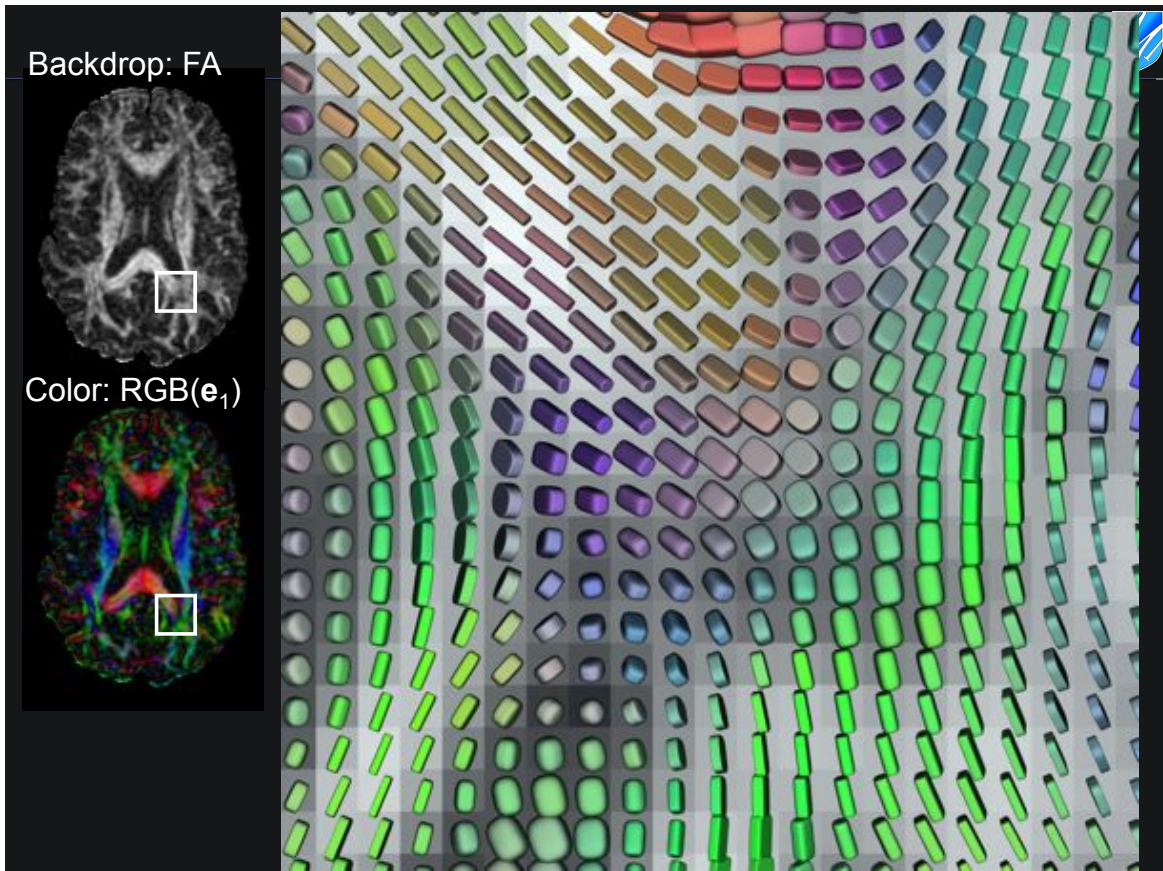
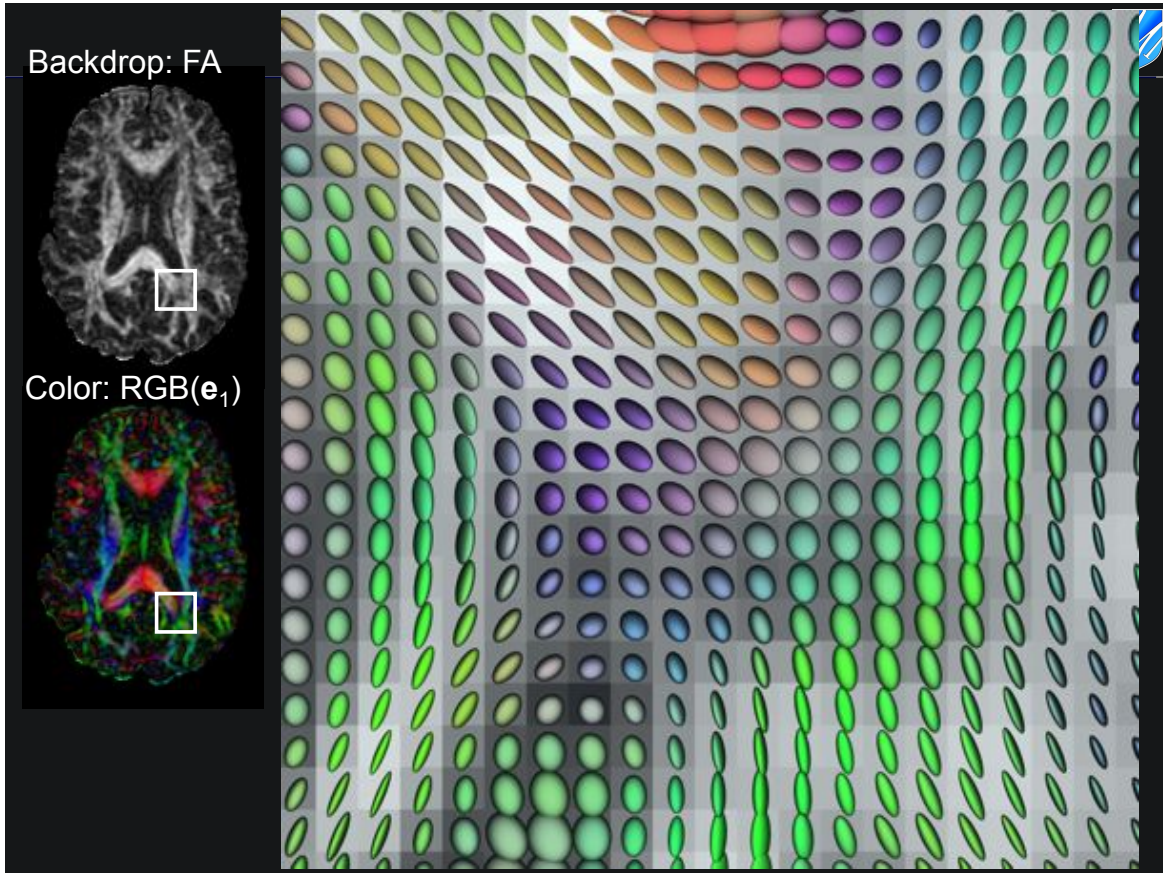


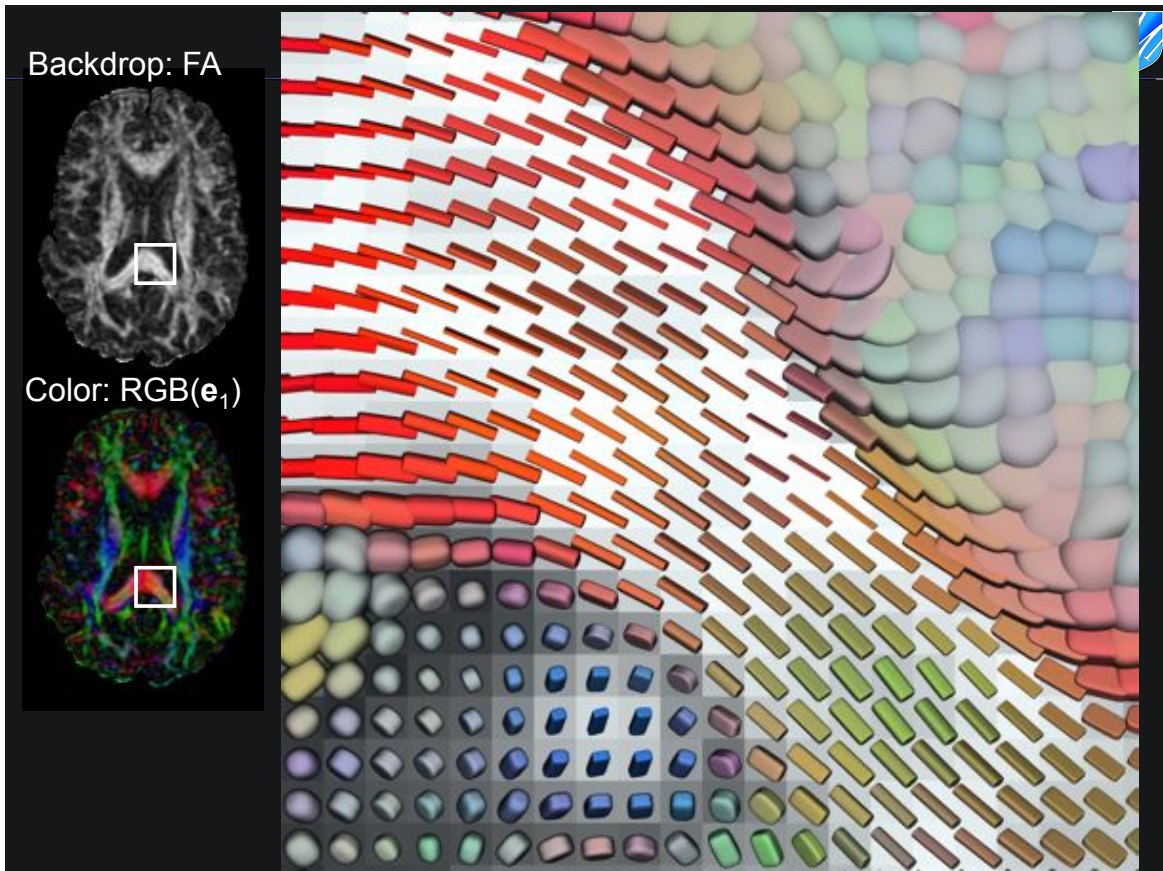
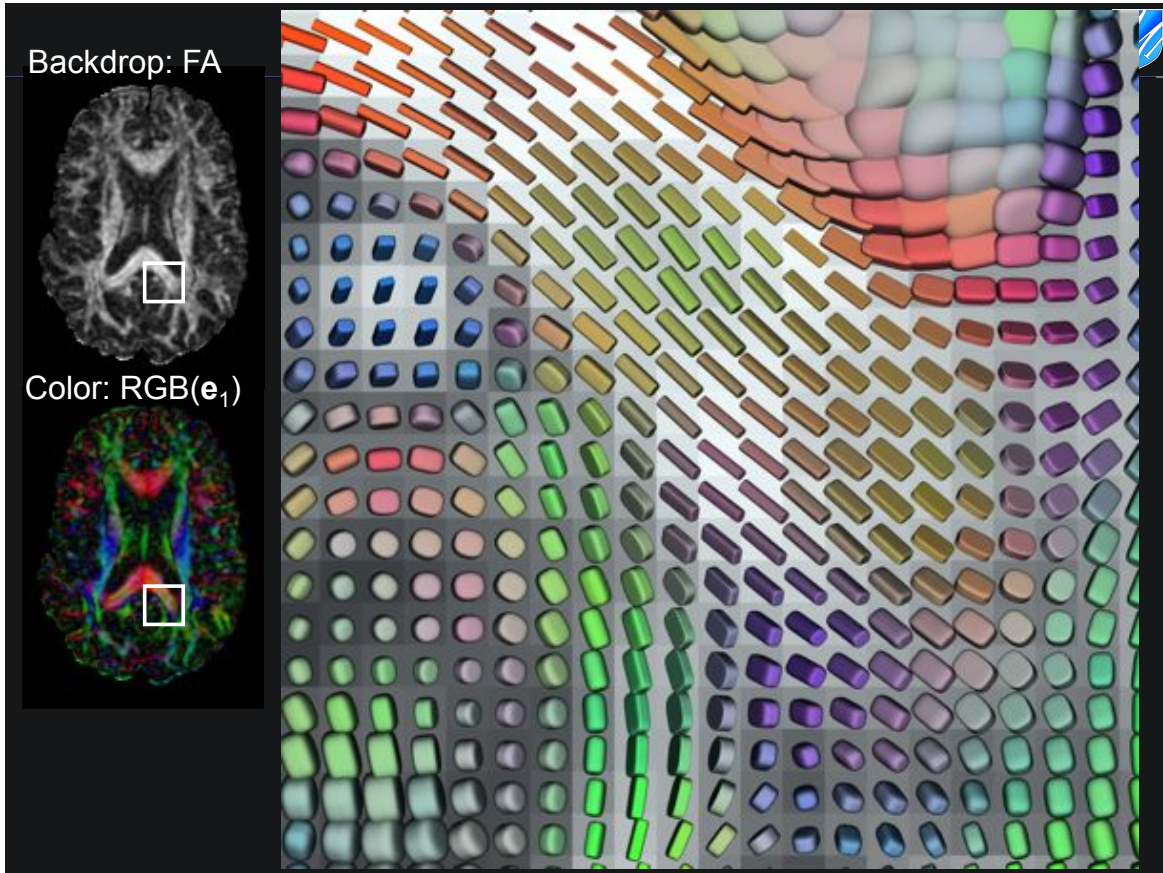
Backdrop: FA

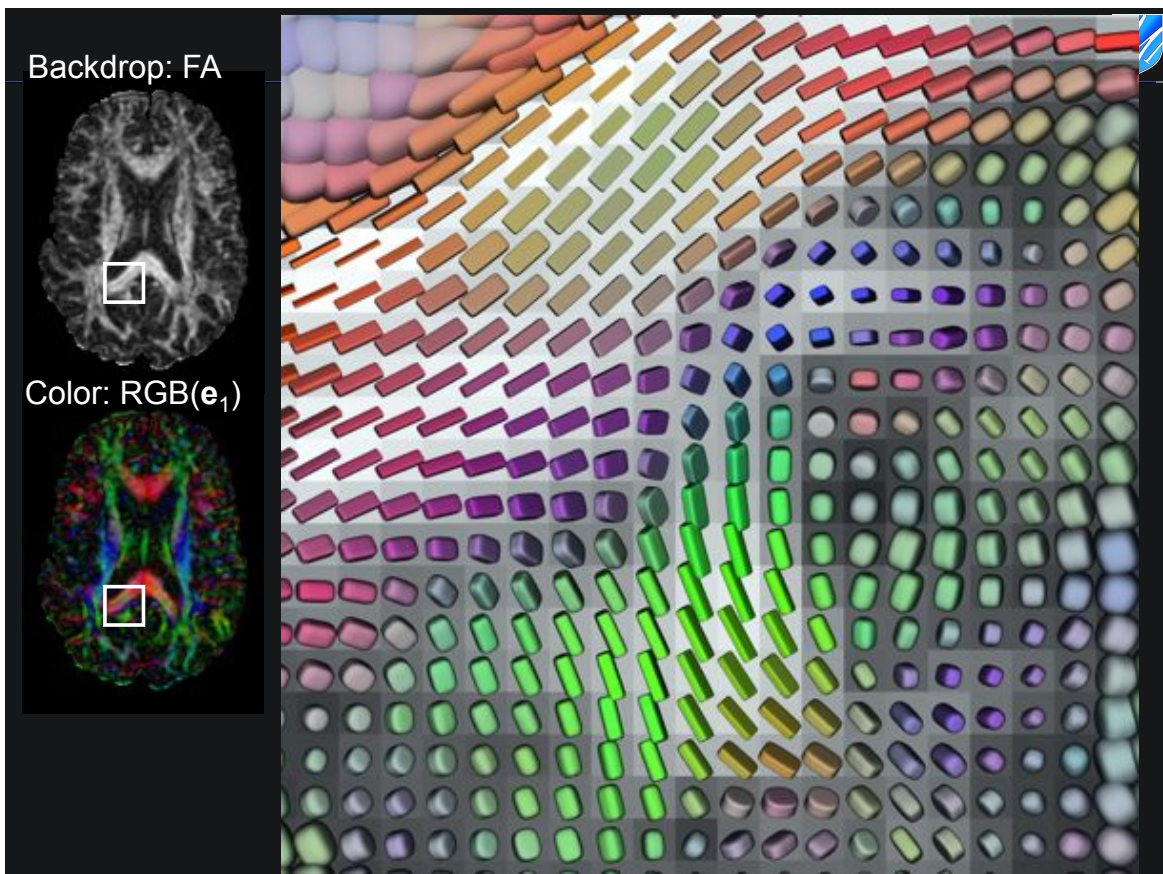
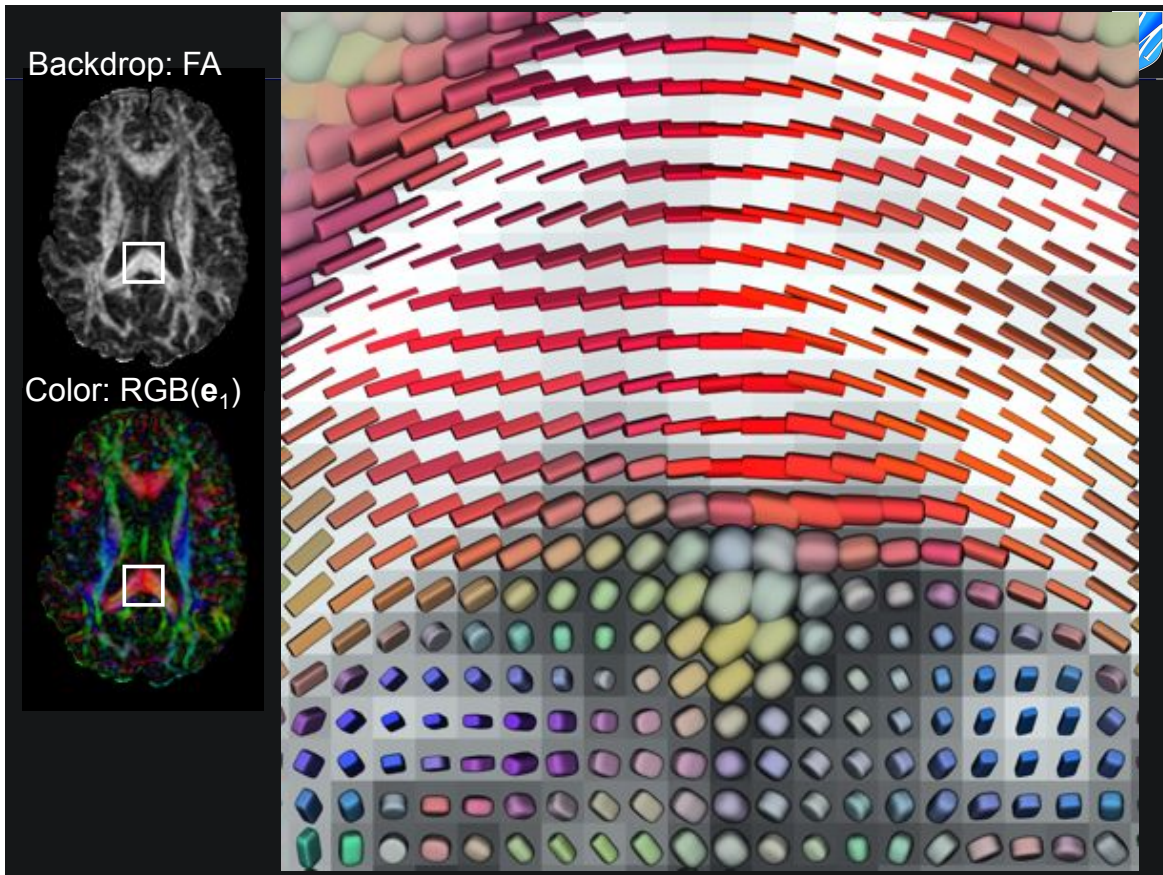


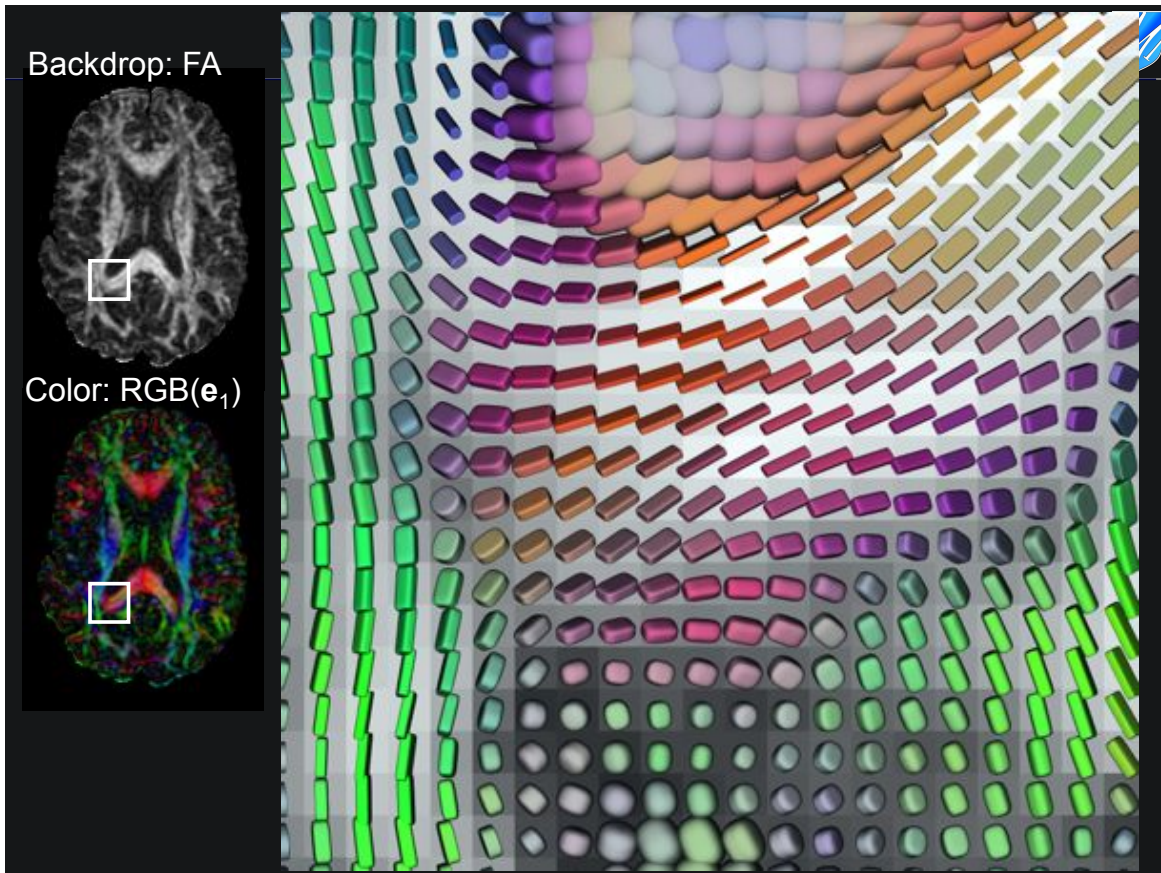
Color: $RGB(e_1)$











Outline

Visualization:

DWI Data source

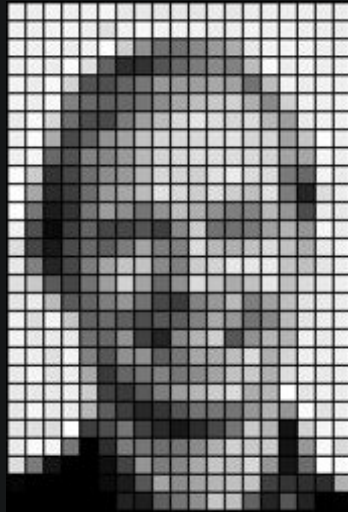
Space of Tensor Shape

Volume Rendering

Superquadric Glyphs

Glyph packing

Appearance matters



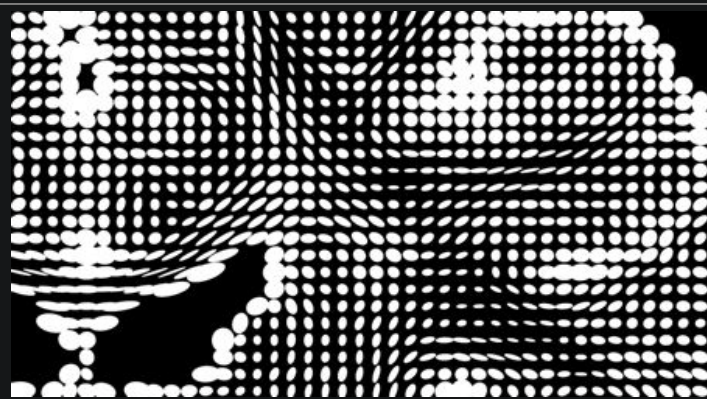
Display method affects ability to perceive data
Don't introduce arbitrary or ill-defined structure

Textures tuned by tensor data, cont.

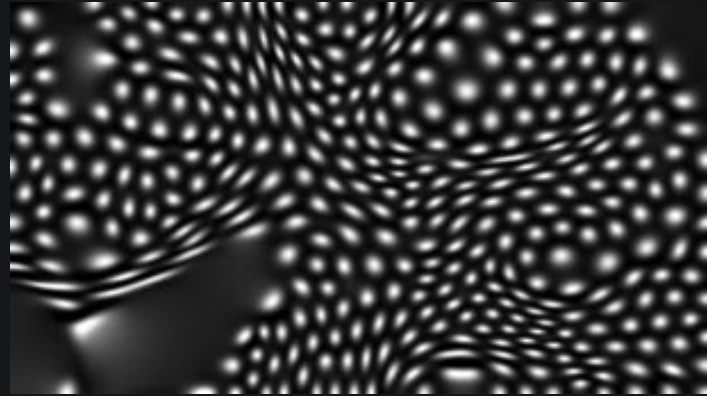


Real data

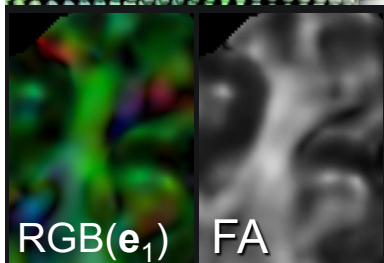
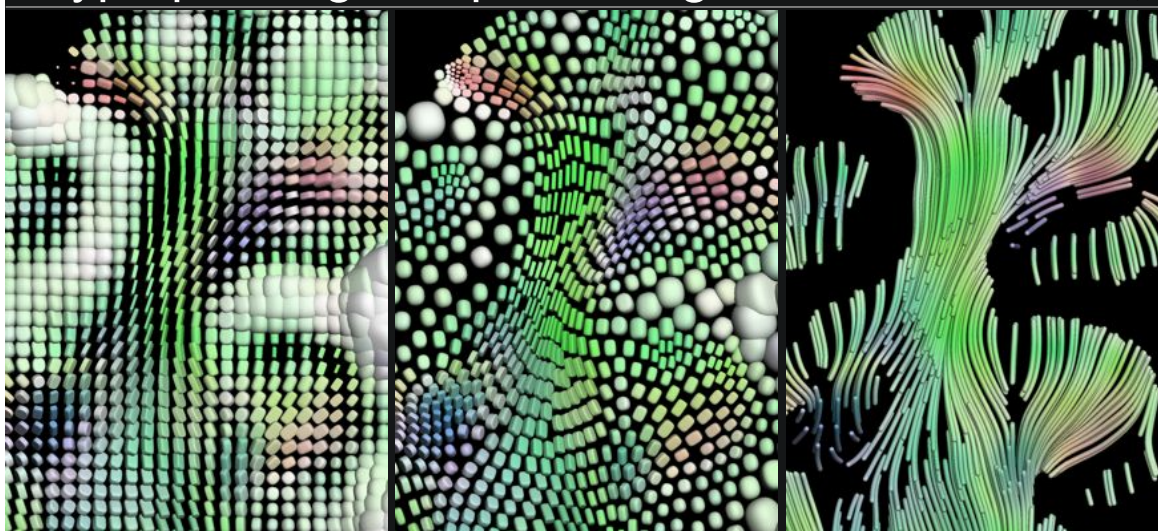
Ellipses



Texture



Glyph packing compared to grid, fibers

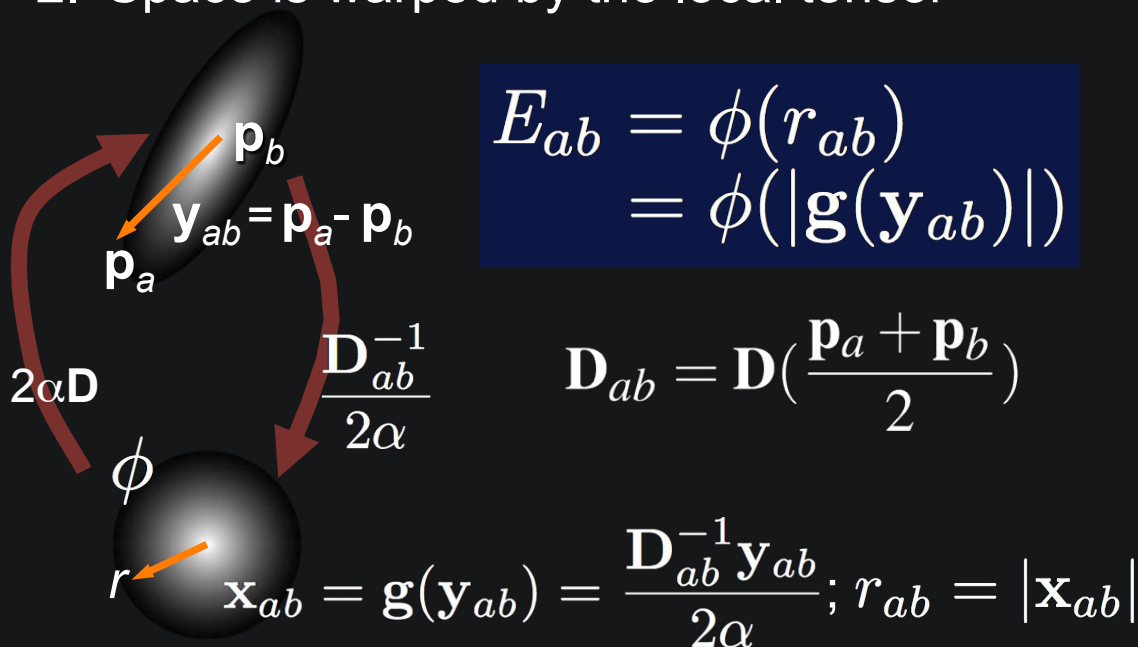


Formation of **discrete** glyphs into a **continuous** texture
Need tensor interpolation

Tensor-based potential energy



1. Particles push each other away
2. Space is warped by the local tensor



Force is spatial derivative of energy



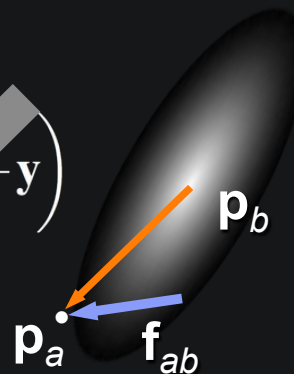
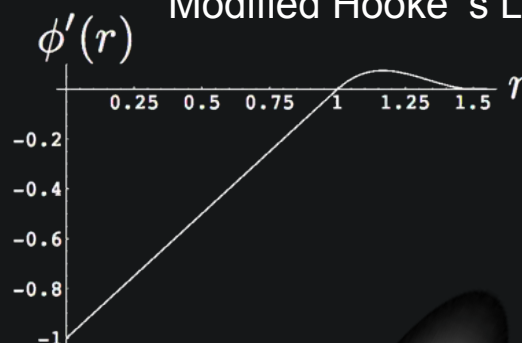
$$E = \phi(|\mathbf{g}(\mathbf{y})|)$$

$$\frac{dE}{d\mathbf{y}} = \frac{dE}{dr} \frac{dr}{d\mathbf{x}} \frac{d\mathbf{x}}{d\mathbf{y}}$$

$$= \phi'(|\mathbf{x}|) \frac{\mathbf{x}^T}{|\mathbf{x}|} \frac{1}{2\alpha} \left(\mathbf{D}^{-1} + \frac{d\mathbf{D}^{-1}}{d\mathbf{y}} \mathbf{y} \right)$$

$$\mathbf{f}_{ab} = -\frac{\phi'(|\mathbf{x}_{ab}|)}{2\alpha} \mathbf{D}_{ab}^{-1} \frac{\mathbf{x}_{ab}}{|\mathbf{x}_{ab}|}$$

Modified Hooke's Law

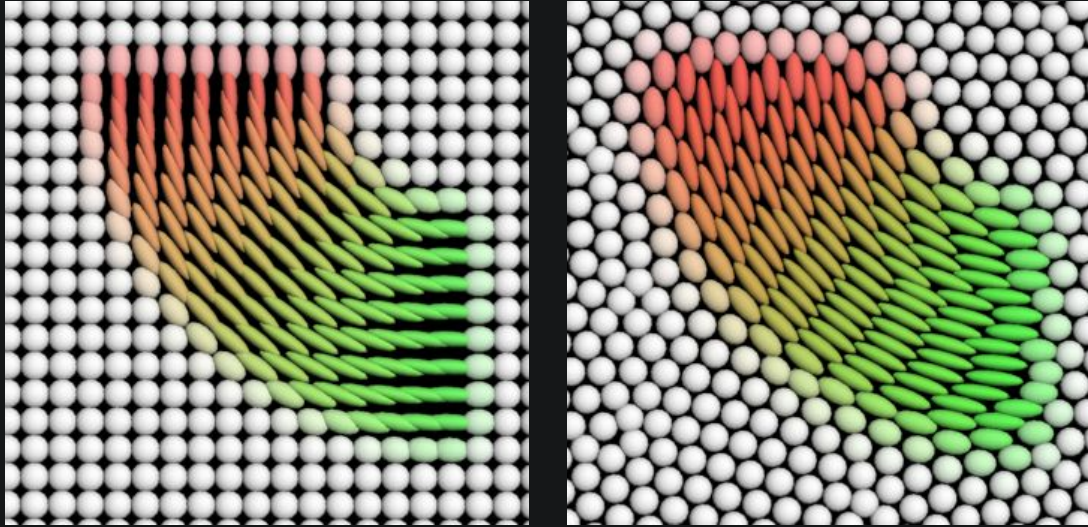


Implementation



- Spatial Binning
- Inverse approximation : $\mathbf{D}_{ab}^{-1} \approx \frac{\mathbf{D}^{-1}(\mathbf{p}_a) + \mathbf{D}^{-1}(\mathbf{p}_b)}{2}$
- Constraints on slices
- Solver: $F=ma$ vs. Gradient descent
 - Order of faster convergence times than paper
- Probabilistically re-use probed tensors

Results: synthetic data



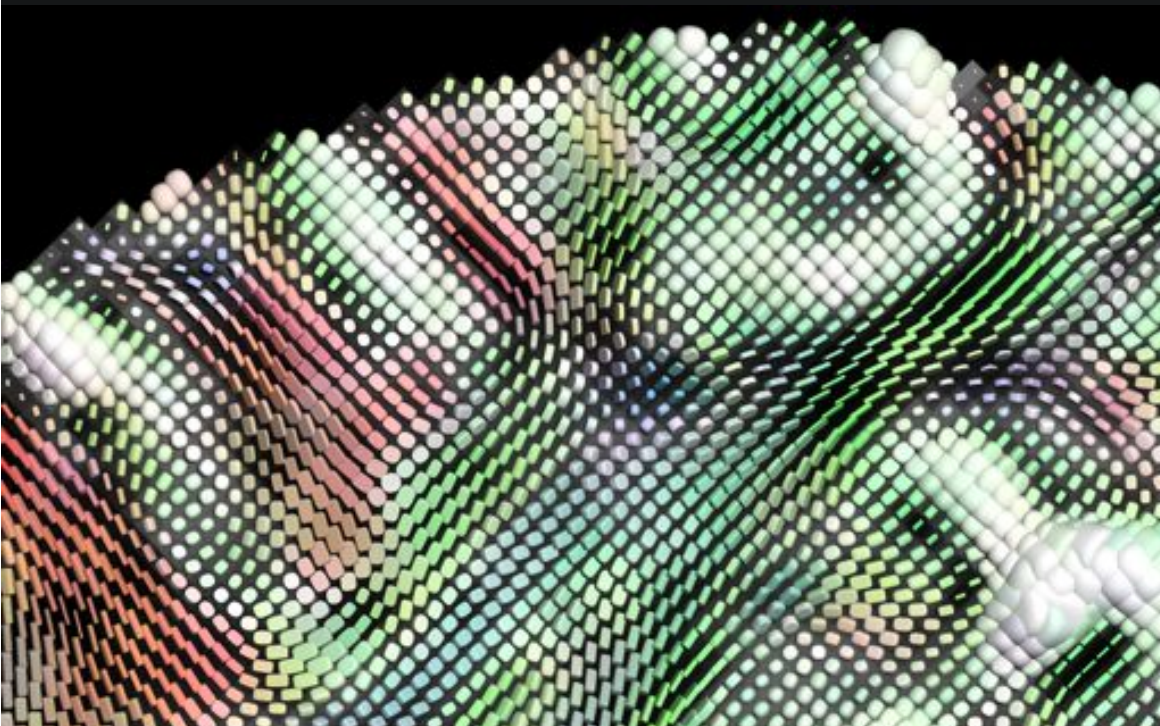
Glyph overlap reduced

Smoother orientation change

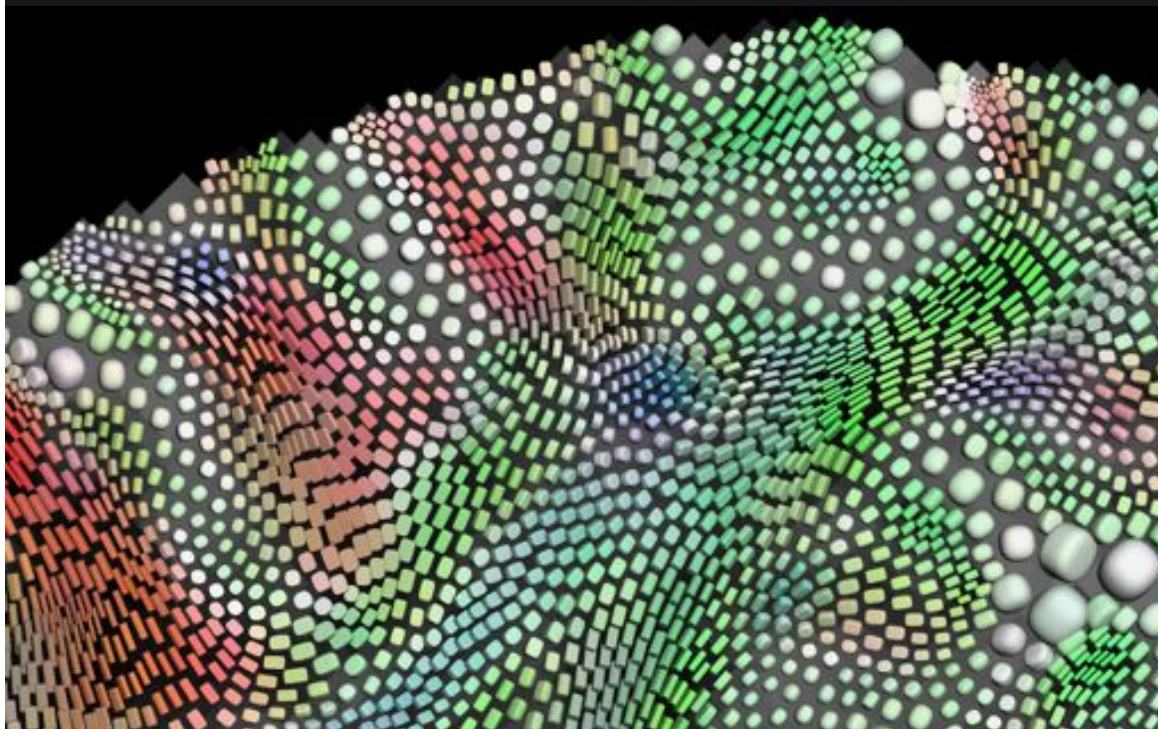
Problem (?): new hexagonal grid

Packing results can seed other glyph geometry

Results: Real data



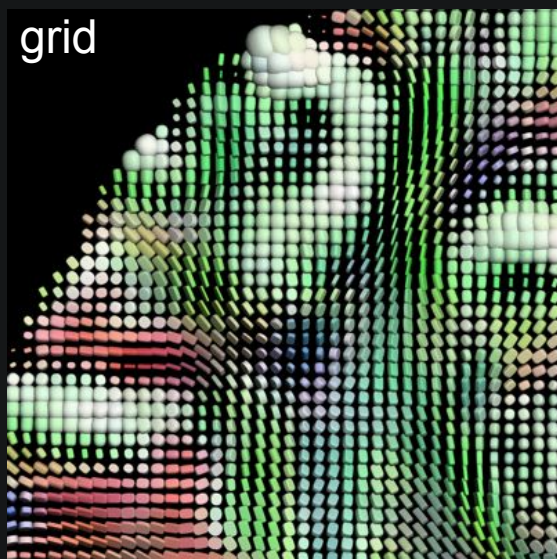
Results: Real data



Results: healthy slice



grid

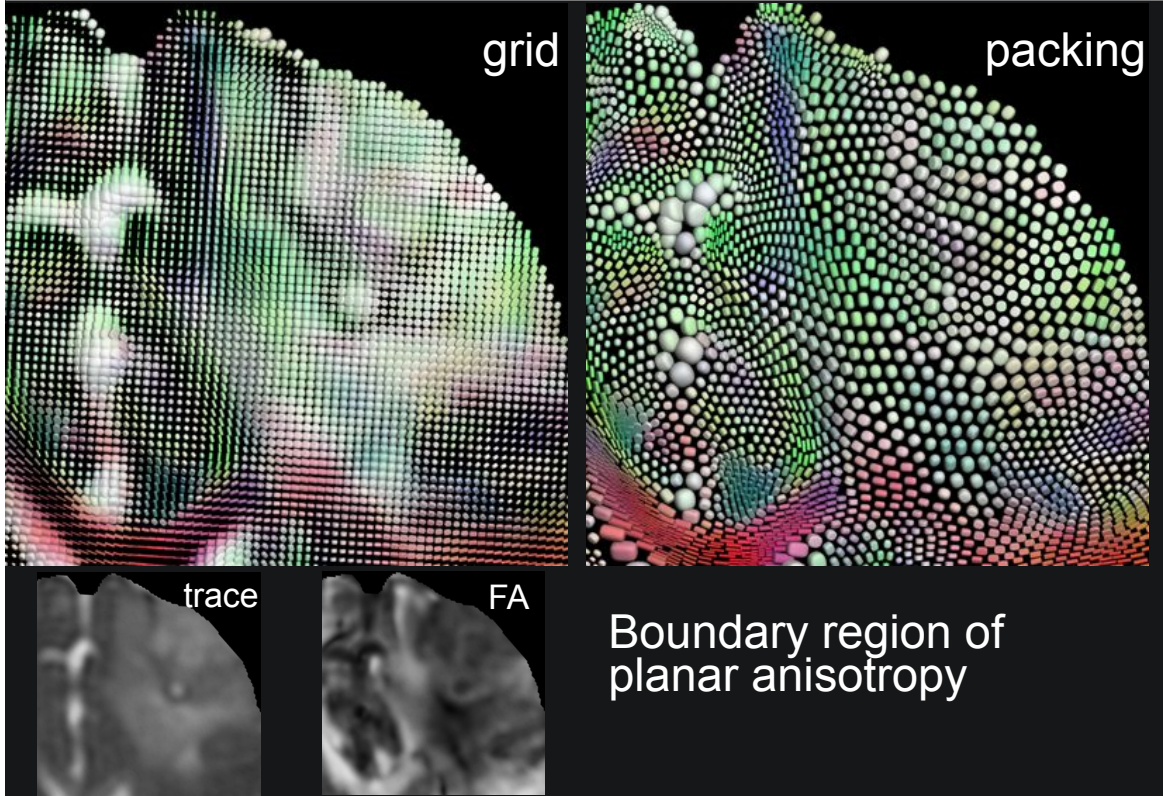


packing



Tracts only in linear anisotropy; glyphs everywhere
Glyph packing complements tractography

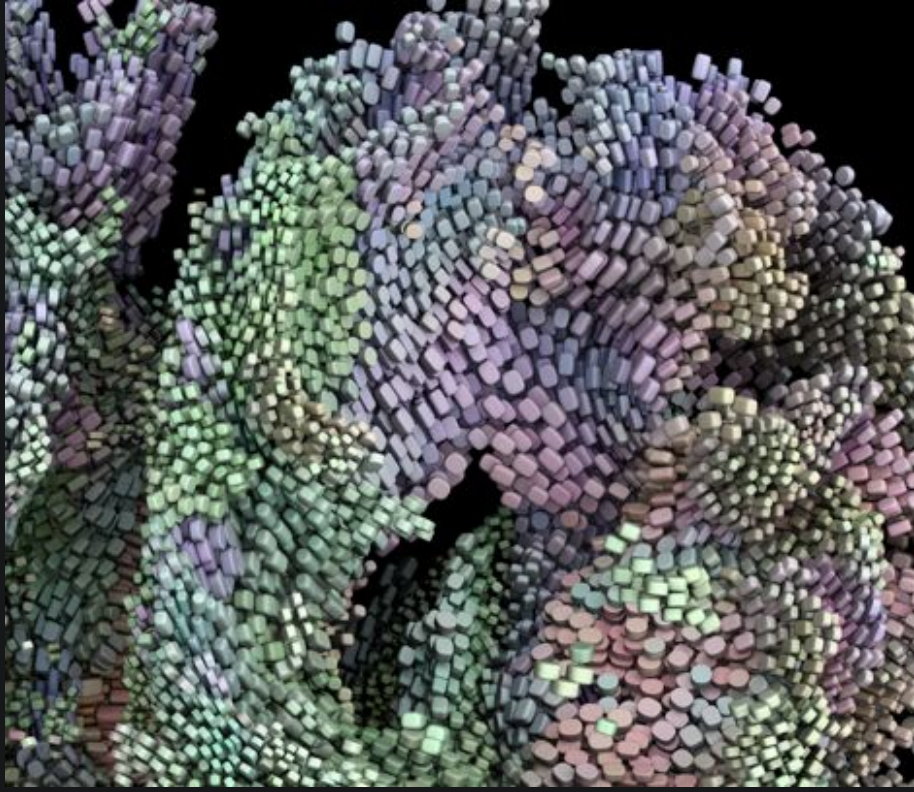
Results: oligodendrogloma slice



Results: volume of glyphs



Results: volume of glyphs



Outline



Software:

Teem basics

Nrrd: for arrays

Gage: for convolution

Ten: for tensors

Deft: interactive



Software:

Teem basics

Nrrd: for arrays

Gage: for convolution

Ten: for tensors

Deft: interactive



Decided in 1999 to never again write
“throw-away” code

Teem == the trail of crumbs of my research

Set of C libraries that play well together

Other people find it useful

Part of SCIRun, Slicer, NA-MIC Toolkit

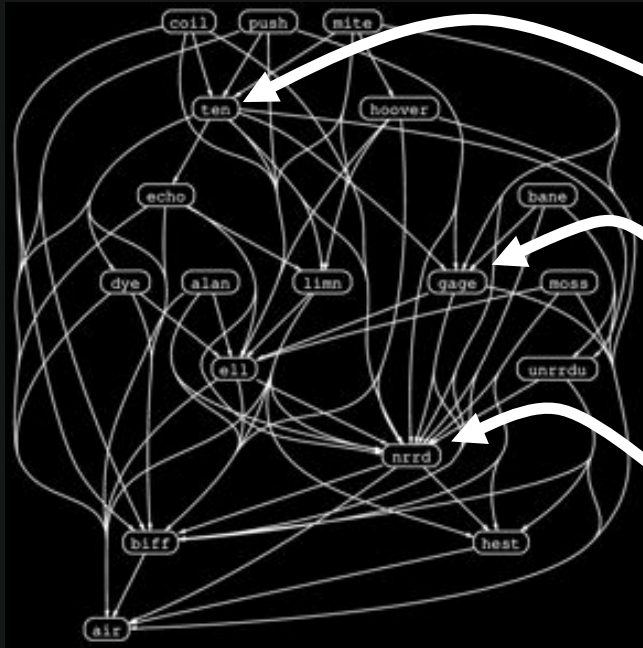
LGPL + Static linking (like FLTK)

<http://teem.sourceforge.net>

Teem



Coordinated set of C libraries for scientific data communication, processing, and visualization



Ten: DWI and DTI processing (tensor estimation, anisotropy metrics, fiber tractography)

Gage: convolution-based point sampling of scalar/vector/tensor fields, and derived attributes

Nrrd: raster data representation, file I/O, and processing

Outline



Software:

Teem basics

Nrrd: for arrays

Gage: for convolution

Ten: for tensors

Deft: interactive



“Nearly Raw Raster Data”

Representation and manipulation of N-dimensional raster data

- File format
- Data structure
- Large set of operations



Associated “unu” command-line tool

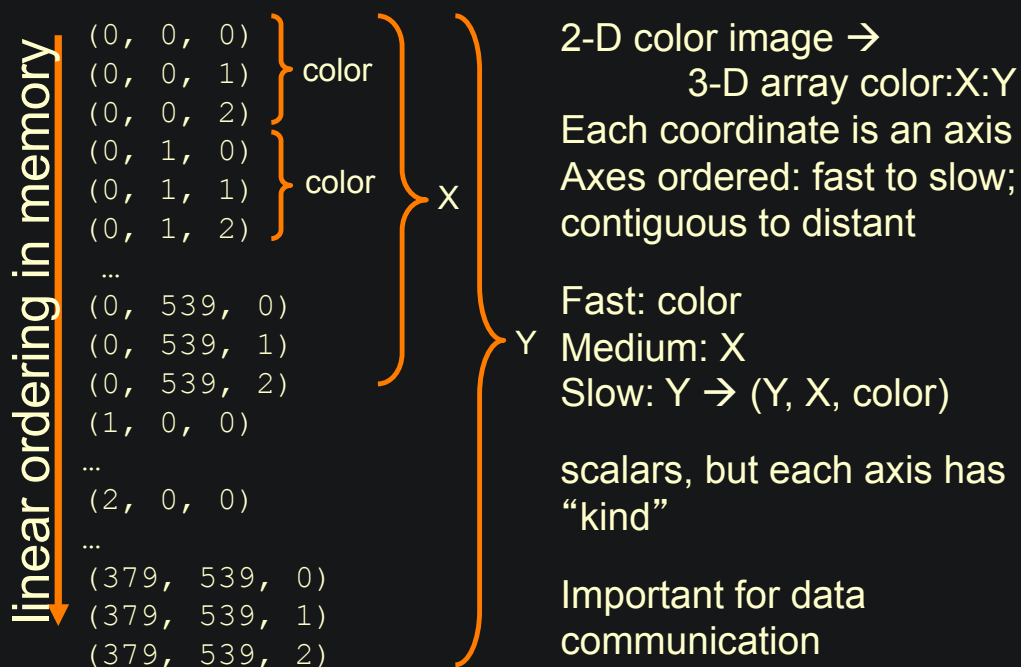
Philosophy: “stark simplicity”

Generality it has is based on experience

Concept: scan-line ordering



N-D raster data has “scan-line ordering”



NRRD DWI headers : Motivation



File format need

- Self-contained representation
- All image values (in original files)
- All DWI-specific parameters (b, g_i , etc.)
- All coordinates (**including measurement frame!!**)
- DICOM is not a solution (DWI or DTI)
 - Manual sign flips/transposition \Rightarrow should **SAVE** answer

NA-MIC solution:

- NRRD **format** + key/value pair **convention**
- NRRD: image values + coordinates
 - **Explicit** representation of measurement frame
- Key/value pair convention: DWI-specific parameters

Now in use in multiple labs at Brigham & Women's, including for clinical neurosurgery

Example NRRD Header



```
NRRD0005
content: NOMIC01
type: short
dimension: 4
space: right-anterior-superior
sizes: 256 256 36 14
thicknesses: NaN NaN 3 NaN
space directions: (-0.9375,0,0) (0,-0.9375,0) (0,0,-3) none
centerings: cell cell cell none
kinds: space space space list
endian: little
encoding: raw
space units: "mm" "mm" "mm"
space origin: (125.00,124.10,79.3)
data file: S4.%03d 1 504 1 2
byte skip: -1
measurement frame: (0,-1,0) (1,0,0) (0,0,-1)
modality:=DWMRI
DWMRI_b-value:=800
DWMRI_gradient_0000:= 0 0 0
DWMRI_NEX_0000:=2
DWMRI_gradient_0002:= -0.8238094 -0.4178235 -0.3830949
DWMRI_gradient_0003:= -0.5681645 0.5019867 -0.6520725
...
```



Volume interleaved, volume at a time:

```
data file: dwi%03d.gipl 1 12 1
byte skip: -1
sizes: 256 256 60 12
kinds: space space space list
```

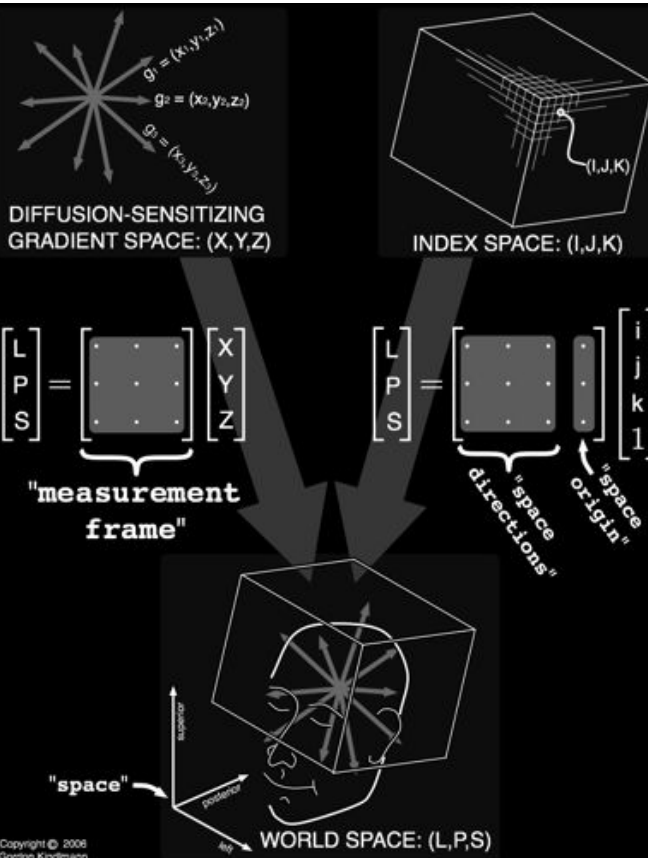
Volume interleaved: slice at a time

```
data file: MR_0028_%04d.dcm 1 448 1 2
byte skip: -1
sizes: 128 128 64 7
kinds: space space space list
```

Slice interleaved: slice at a time

```
data file: test_%04d.dcm 1 448 1 2
byte skip: -1
sizes: 128 128 7 64
kinds: space space list space
```

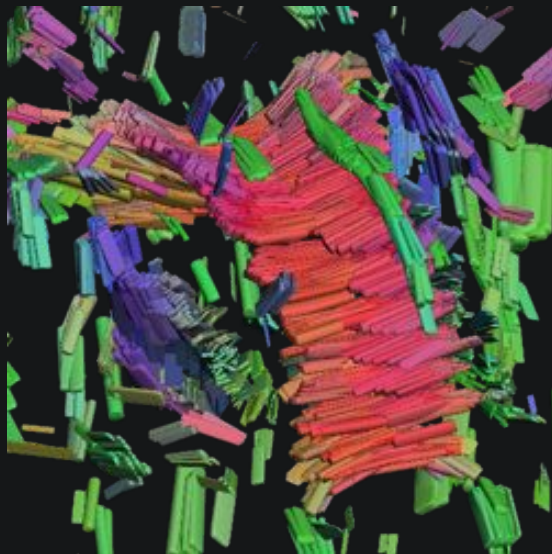
NRRD's full representation of image and data orientation



Measurement frame mysteries

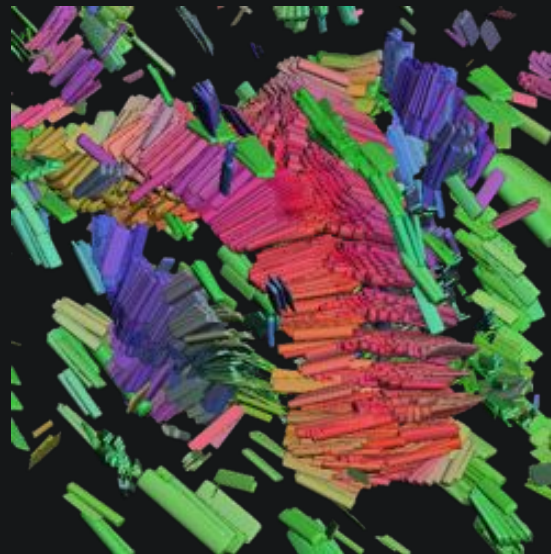


Maps from gradient (X,Y,Z) to world (R,A,S)/(L,P,S)



measurement frame:

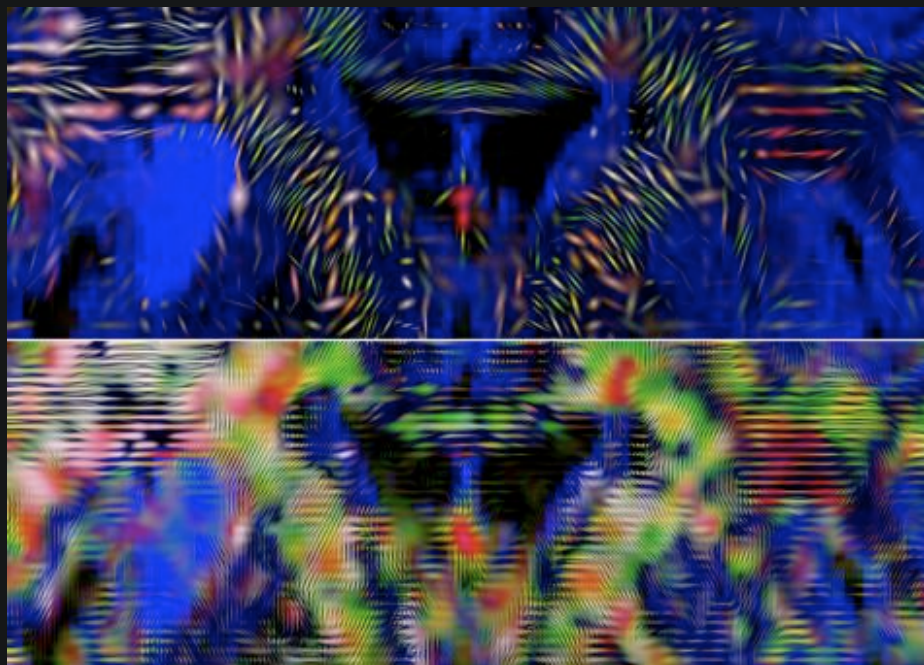
$(0,-1,0)$ $(1,0,0)$ $(0,0,-1)$



measurement frame:

$(0,1,0)$ $(1,0,0)$ $(0,0,1)$

Measurement frame bugs in pubs



Benger '06

Outline



Software:

Teem basics

Nrrd: for arrays

Gage: for convolution

Ten: for tensors

Deft: interactive

Gage



Convolution-based measurements

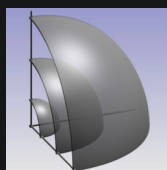
Discrete data, continuous kernels

Scalars, 3-Vectors, DTIs, DWIs

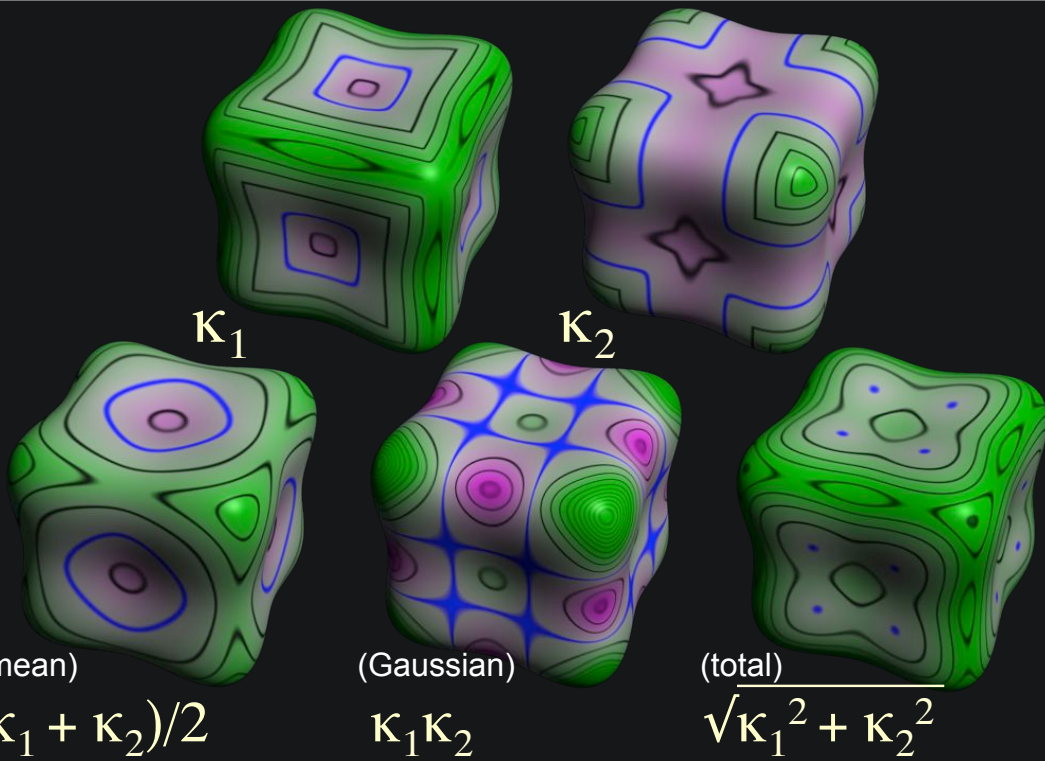
Values, 1st, 2nd derivatives

e.g. curvature tensor in scalars

Basis of volume rendering, tractography



Curvature measures



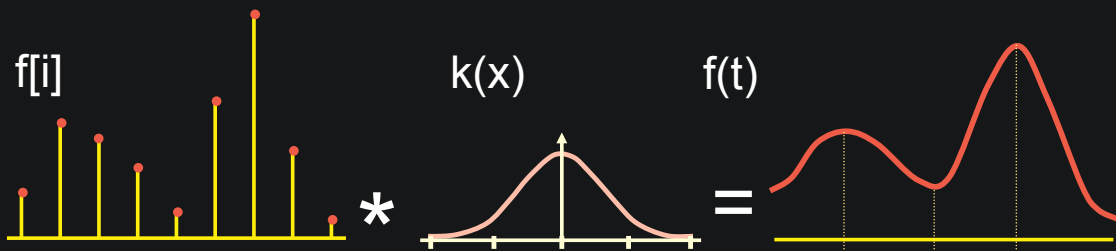
Volume NPR: results



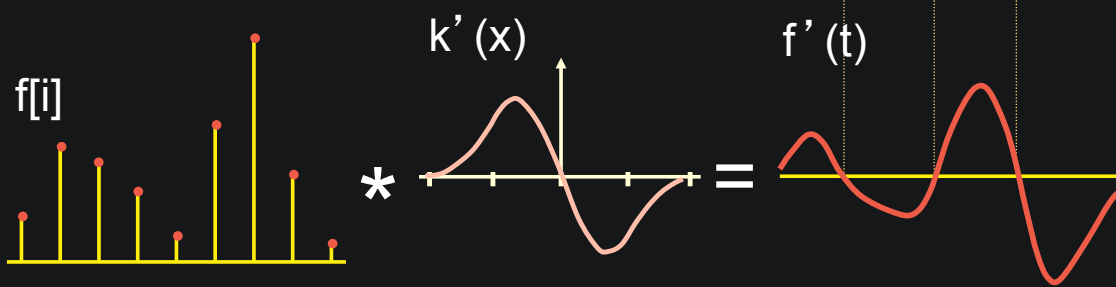
Measurement (of scalars) by convolution



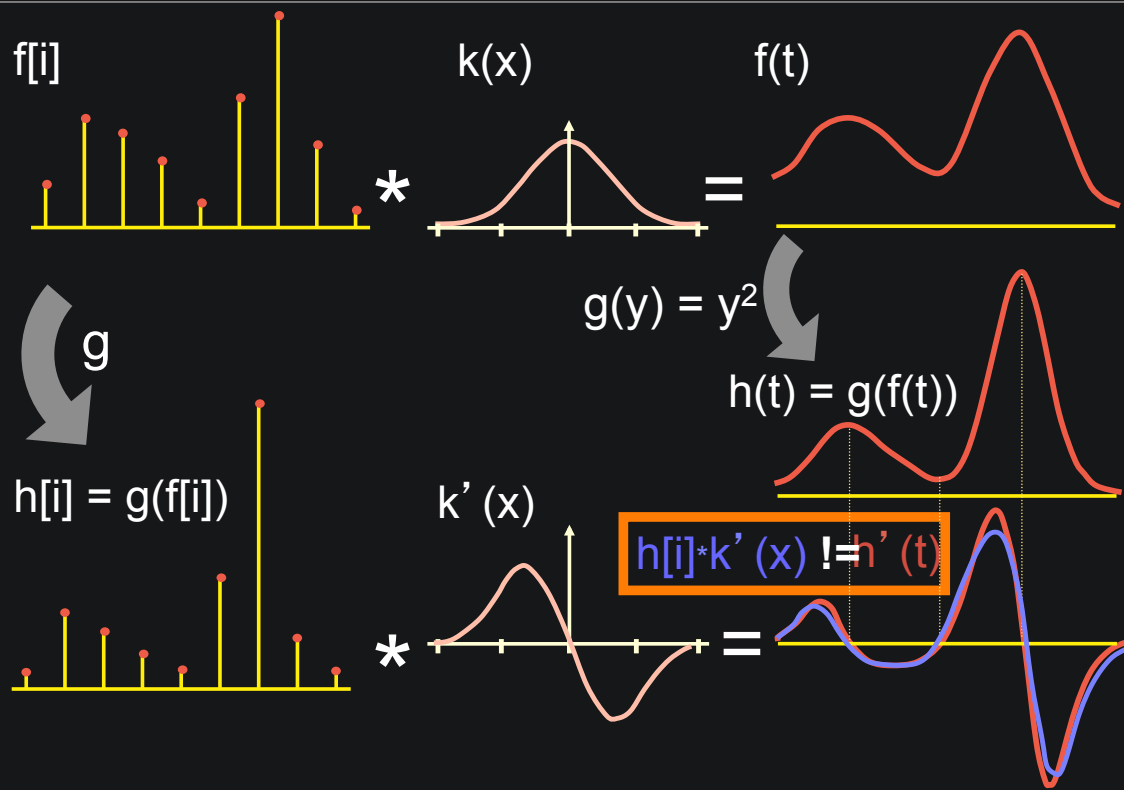
Continuous field: convolution of sampled coefficients with continuous reconstruction kernels



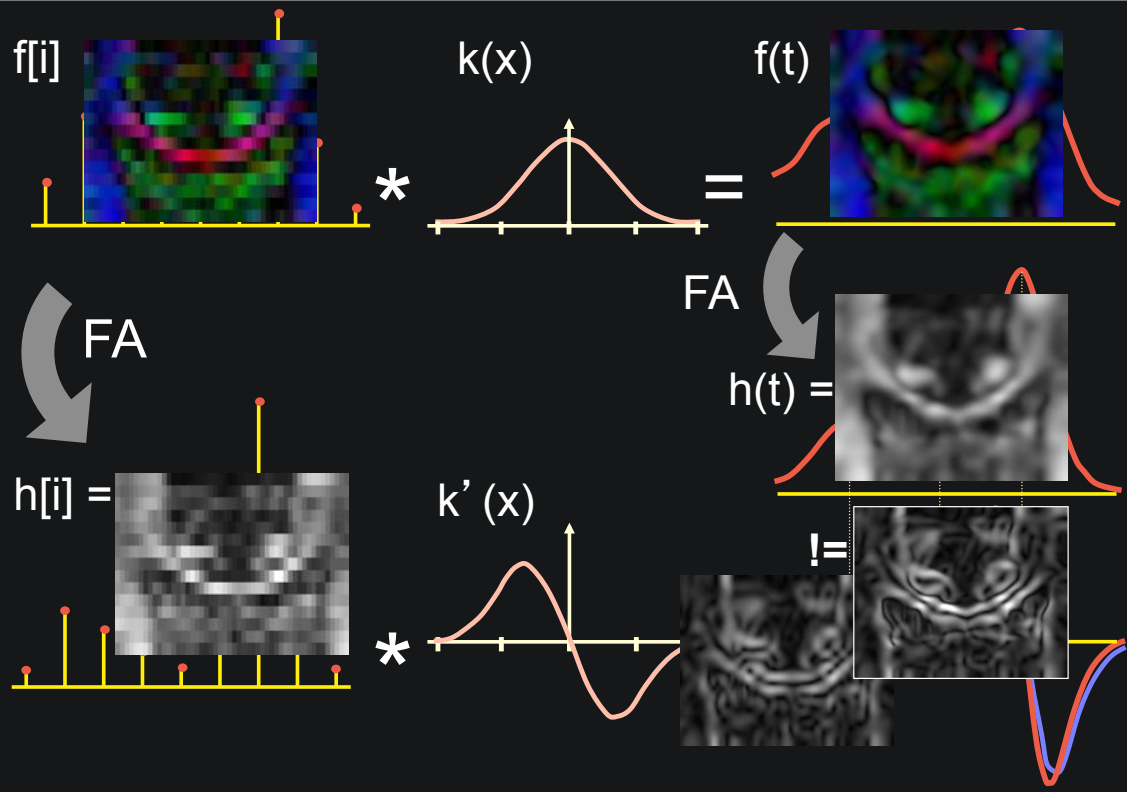
Differentiation: convolve w/ derivative of reconstruction kernel



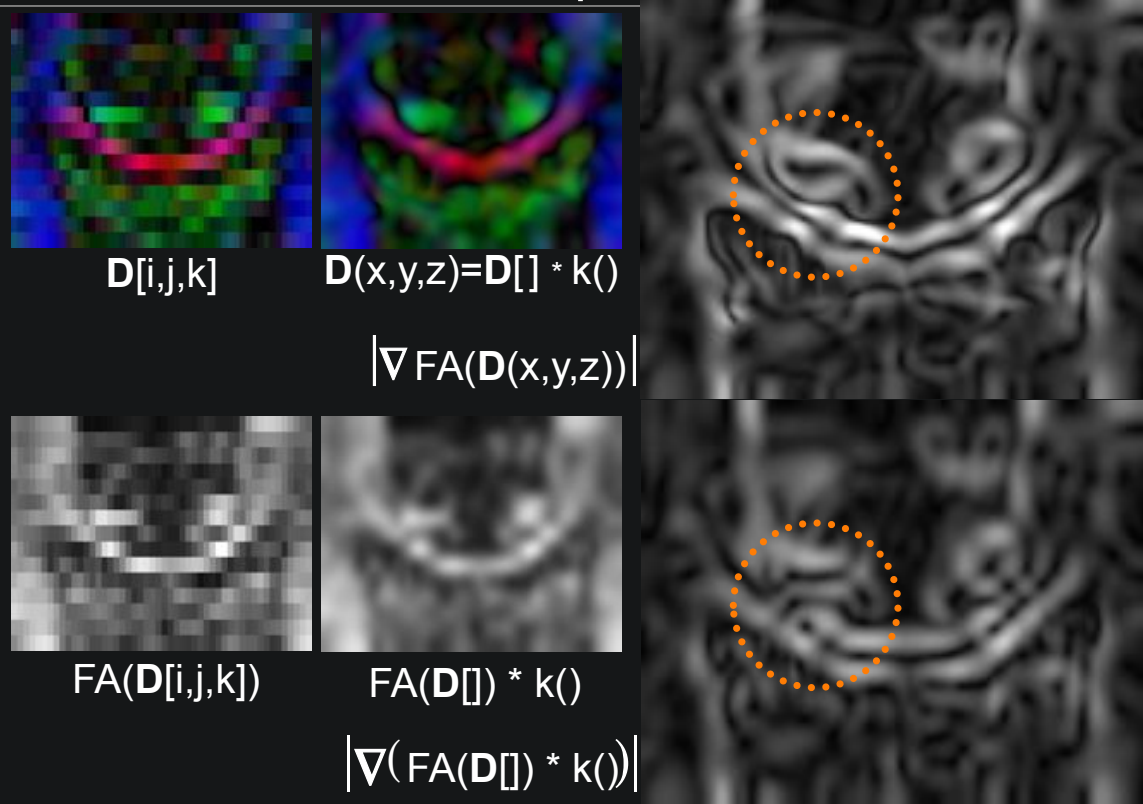
Non-linear transform of data



Fractional Anisotropy (FA) is non-linear



FA is non-linear, close-up



FA from invariants, from coefficients



$$\begin{aligned}
 \text{FA} &\equiv \sqrt{\frac{3 \mathbf{D}:\mathbf{D}}{2 \mathbf{D}:\mathbf{D}}} & D &= \mathbf{D} - \langle \mathbf{D} \rangle \mathbf{I} & \mathbf{D}:\mathbf{D} &= \text{tr}(\mathbf{D}\mathbf{D}^T) \\
 \text{FA} &= 3\sqrt{\frac{Q}{S}} & Q &= \frac{S - J_2}{9} & J_2 &= D_{xx}D_{yy} + D_{xx}D_{zz} + D_{yy}D_{zz} \\
 & & & & & - D_{xy}^2 - D_{xz}^2 - D_{yz}^2 \\
 & & & & S &= \mathbf{D}:\mathbf{D} = D_{xx}^2 + D_{yy}^2 + D_{zz}^2 \\
 & & & & & + 2D_{xy}^2 + 2D_{xz}^2 + 2D_{yz}^2 \\
 \nabla Q &= \frac{\nabla S - \nabla J_2}{9} & \nabla J_2 &= (D_{yy} + D_{zz})\nabla D_{xx} + (D_{xx} + D_{zz})\nabla D_{yy} + (D_{xx} + D_{yy})\nabla D_{zz} \\
 & & & - 2D_{xy}\nabla D_{xy} - 2D_{xz}\nabla D_{xz} - 2D_{yz}\nabla D_{yz} \\
 \nabla \text{FA} &= \frac{3}{2} \left(\sqrt{\frac{1}{SQ}} \nabla Q - \sqrt{\frac{Q}{S^3}} \nabla S \right) & \nabla S &= 2D_{xx}\nabla D_{xx} + 2D_{yy}\nabla D_{yy} + 2D_{zz}\nabla D_{zz} \\
 & & & + 4D_{xy}\nabla D_{xy} + 4D_{xz}\nabla D_{xz} + 4D_{yz}\nabla D_{yz}
 \end{aligned}$$

Hessian(FA) more complicated, but similarly derived

Outline



Software:

Teem basics

Nrrd: for arrays

Gage: for convolution

Ten: for tensors

Deft: interactive

Ten



All things tensor-related:

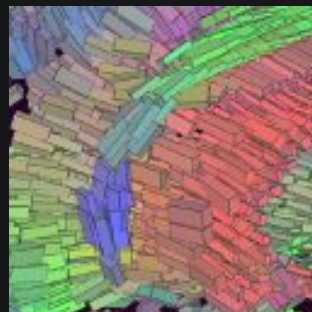
DWI gradient optimization

Tensor estimation

Anisotropy calculations

Fast 3x3 eigensolution

Tractography



Outline



Software:

Teem basics

Nrrd: for arrays

Gage: for convolution

Ten: for tensors

Deft: interactive

Deft: interactive tensor vis



Deft: written in C++/OpenGL

- Built on top of Teem

Compiled with Cmake

MIT License

Deft is **not** for clinicians

low-level data inspection: “xv for tensors” (glyphs)

Triangle strips, glyph palette

Algorithmic visualization

Tractography parameter space



Esto es Todo

Gracias

Preguntas, por
favor?