

Tensor field visualization – My view on the field

January 2016

Geilo winter School – Scientific Visualization

Ingrid Hotz – Linköping University



Tensor field visualization – My view on the field

From the general perspective ...



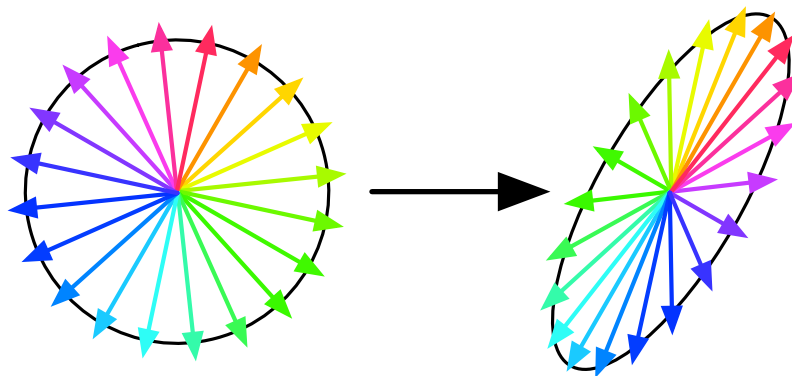
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... to digging in the sand



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Tensor field visualization – My view on the field

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Tensors – Why Do We Care?

- Tensors are everywhere (Simulations, physical theories)
- Can hardly be seen anywhere – nobody cares?


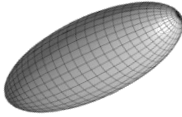
Tensors – Why Do We Care?

If you're sitting at a cocktail party with a bunch of engineers, physicists, and mathematicians, and you want to start a heated debate,

Just ask out loud: “What is a tensor?”

- One person will say that, for all practical purposes, a tensor is just a **fancy word for a matrix**.
- Then someone else will pipe up indignantly and insist that a tensor is a **linear transformation from vectors to vectors**.
- Yet another person will say that a tensor is an ordered **set of numbers that transform in a particular way upon a change of basis**.
- Other folks (like us) will start babbling about **“dyads” and “dyadics.”**

Tensors – Why Do We Care?

<ul style="list-style-type: none">• Vector visualization	<ul style="list-style-type: none">• Tensor visualization
<ul style="list-style-type: none">• Intuitive notion of a vector	<ul style="list-style-type: none">• Intuitive notion?
	
<ul style="list-style-type: none">• Prevalent Applications	<ul style="list-style-type: none">• Many different applications
<ul style="list-style-type: none">• Some driving questions	<ul style="list-style-type: none">• Only a few shared questions
<ul style="list-style-type: none">• Long visualization history, well accepted	<ul style="list-style-type: none">• No visualization history, hardly known

Tensors – Why Do We Care?

Data ??

Head



Computer

No common language across applications
But terminology mostly related to **specific tensor invariants**
→ Basis of our visualization concept

Eigenvalues
Eigenvectors

Questions

Fuzzy



Crisp

Often scientists have no glue how to expect from the tensor data

Tensors – Why Do We Care?

Tensor visualization is field that is still in its infancy

(maybe except for Diffusion Tensor Imaging

however this does not really happen in the visualization community)

Overview

I. Tensors

As mathematical objects

As physical descriptors

II. Some basic visualization methods

III. Tensor Invariants for feature definition

IV. The story of a collaboration

I. Tensors as mathematical objects

Tensor field visualization – My view on the field

I Tensors as mathematical objects

V n -dim vector space

Second-order tensor T
is a bilinear function

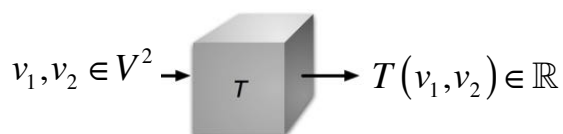
$$T : V \otimes V \rightarrow \mathbb{R}$$

$$T(v, w) = w^T \cdot \mathbf{T} \cdot v, \quad \forall v, w \in V, \mathbf{T} \in \mathbb{R}^{n \times n}$$

Second-order tensor T
as lin. operator

$$T : V \rightarrow V$$

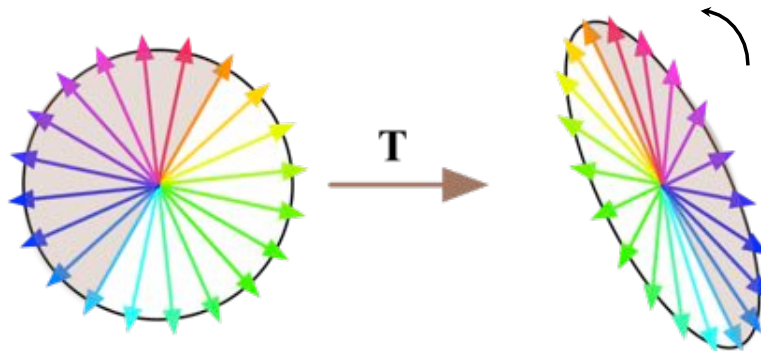
$$T(v) = \mathbf{T} \cdot v, \quad \forall v \in V$$



$$\mathbf{T} = \begin{pmatrix} t_{11} & \cdots & t_{1n} \\ \vdots & \ddots & \vdots \\ t_{n1} & \cdots & t_{nn} \end{pmatrix}$$

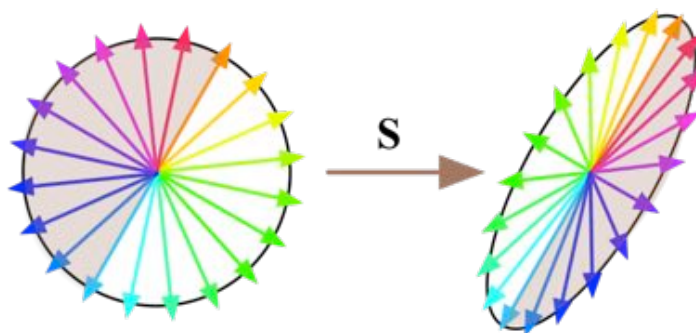
This is not a tensor but a matrix
Often a tensor is mixed up with
its representation

I Tensors as mathematical objects



The tensor is uniquely determined by its action on all unit vectors
Rotation, deformation, reflection

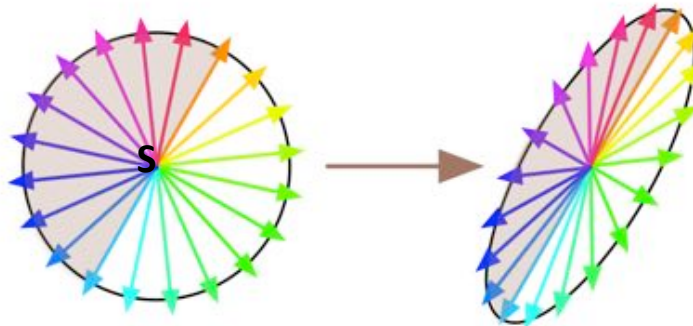
I Tensors as mathematical objects



Symmetric tensor

Deformation, reflection – no rotation

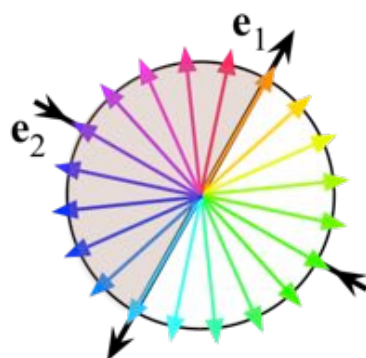
I Tensors as mathematical objects



Positive definite, symmetric tensor

Deformation – no rotation, no reflection

I Tensors as mathematical objects



- Tensors can be analyzed using any convenient reference frame
 - For specific reference frames, the tensor representation becomes especially simple
- Basis given by the eigenvectors of the matrix

Symmetric tensor

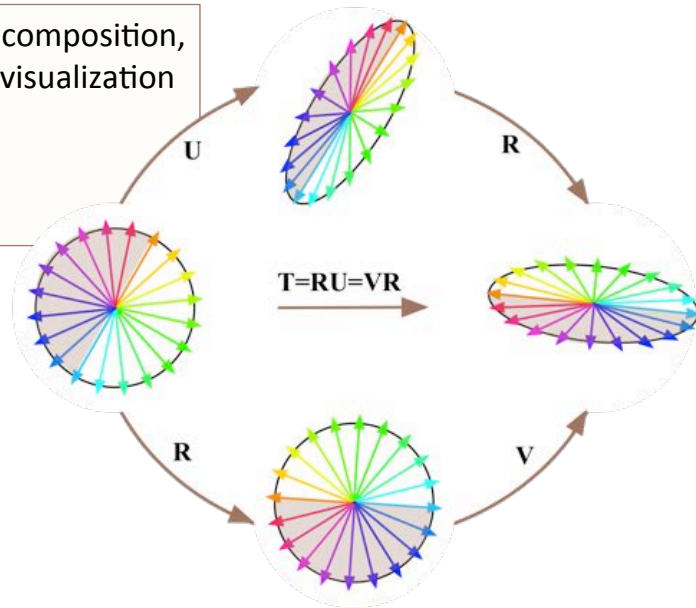
Eigenvectors and eigenvalues

$$T e_i = \lambda_i e_i$$

I Tensors as mathematical objects

There are different tensor Decomposition, which are often the basis for visualization methods

- Polar decomposition – stretch-rotation



Concatenation of mappings

I Tensors as mathematical objects

There are different tensor Decomposition, which are often the basis for visualization methods

- Symmetric and anti-symmetric part

Sum of mappings

$$\mathbf{S} = \frac{1}{2}(\mathbf{T} + \mathbf{T}^T),$$

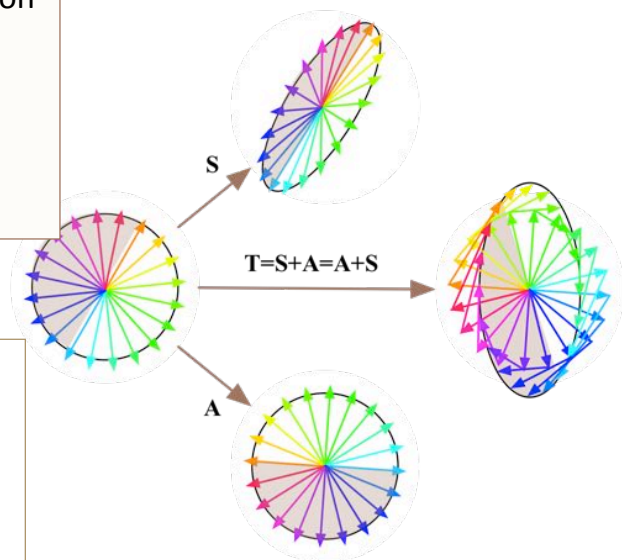
$$s_{ij} = \frac{1}{2}(t_{ij} + t_{ji}).$$

Symmetric part

$$\mathbf{A} = \frac{1}{2}(\mathbf{T} - \mathbf{T}^T),$$

$$a_{ij} = \frac{1}{2}(t_{ij} - t_{ji}),$$

Anti-symmetric part

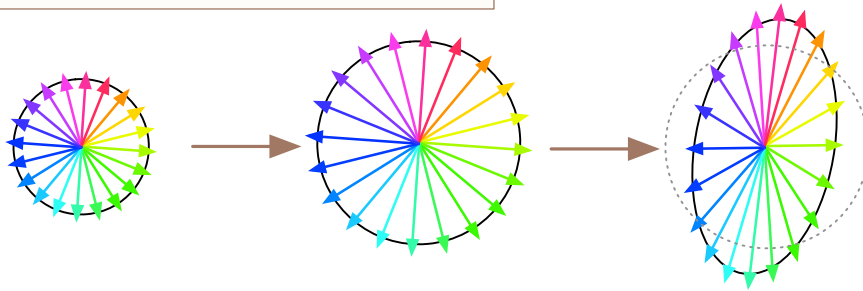


I Tensors as mathematical objects

There are different tensor Decomposition, which are often the basis for visualization methods

- Isotropic scaling, anisotropic deformation (deviator)

- Many different measures for anisotropy available



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I Tensors as mathematical objects

Tensor Invariants

Entities that do not depend on the representation

Properties inherent to the tensor

Examples

- Eigenvalues
- Determinant
- Trance
- All functions that only depend on the eigenvalues

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I Tensor as physical descriptors

Tensor field visualization – My view on the field

I Tensor as physical descriptors

Tensors are everywhere in Physics and Engineering

- Not because the world is linear
- But because linear is simple

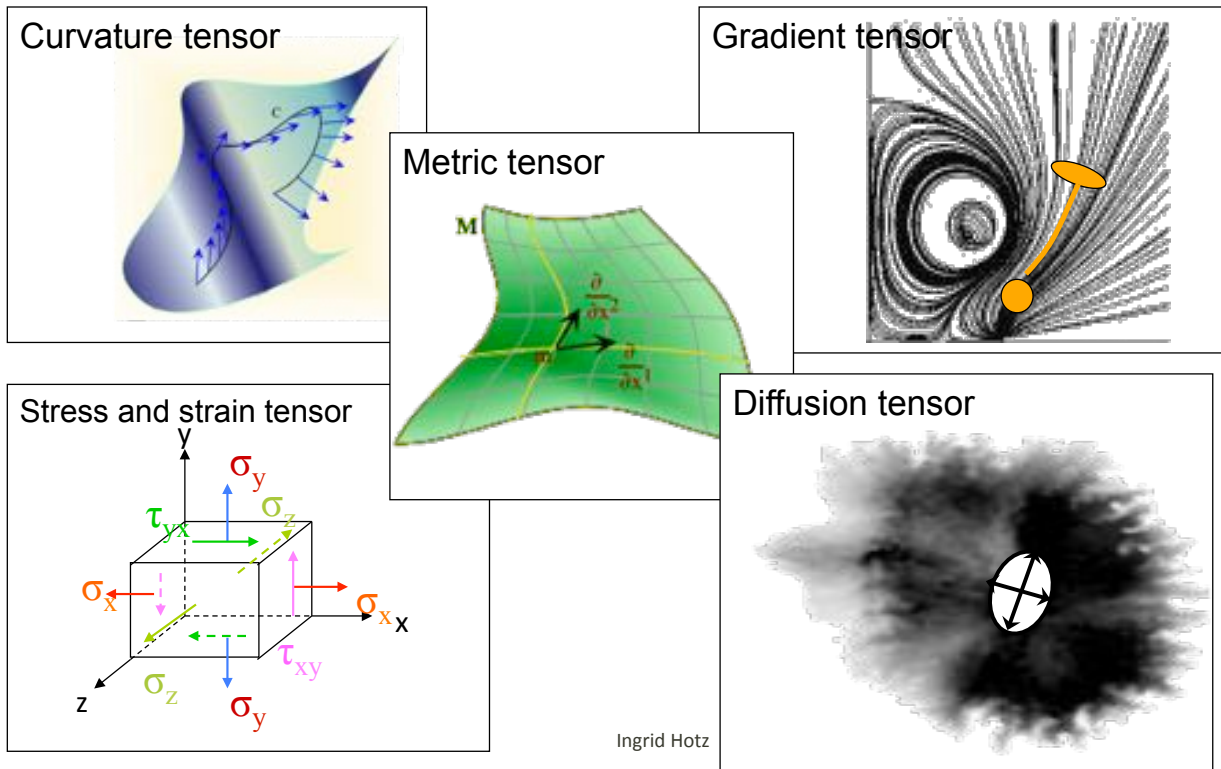
First order descriptors of the dependence of two vector fields v, w

→ first term of the Taylor expansion

$$v(w) = \mathbf{T} \cdot w + \text{higher order terms}$$

They provide a more or less good approximation of the reality

I Tensor as physical descriptors

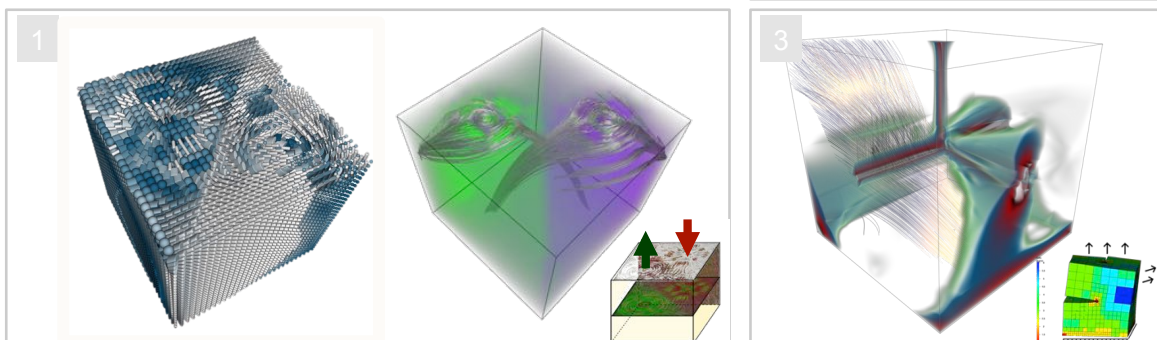


I Tensor as physical descriptors

Stress Tensor

Geo- and Material Sciences

1. Solid block, with two applied forces
2. Implant design, stress simulation in human bone
3. Notched block with external forces



Images: Kratz, ZIB

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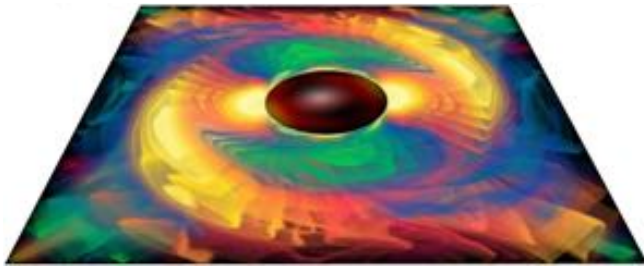
Tensor field visualization – My view on the field

I Tensor as physical descriptors

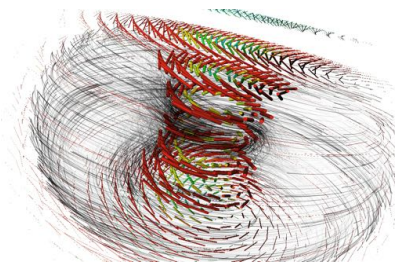
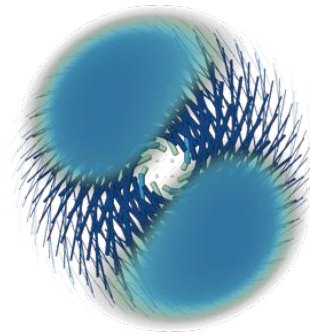
Metric, Curvature, Stress

General Relativity

- Simulation of gravitational field of a rotating black hole, respectively neutron star



Images: Bengner, Kratz, ZIB



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Tensor field visualization – My view on the field

I Tensor as physical descriptors

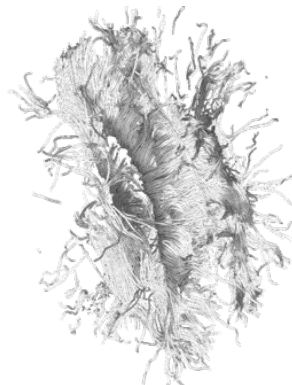
Diffusion Tensor – Medicine

- Imaging method: based on magnetic resonance tomography (MRT) measuring the diffusion of water molecules in tissues
- Application: Reconstruction of neural fibers in human brain (tractography)



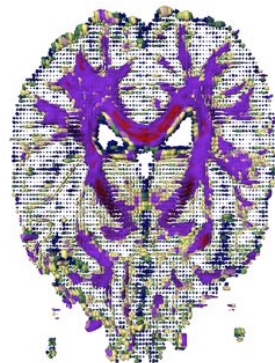
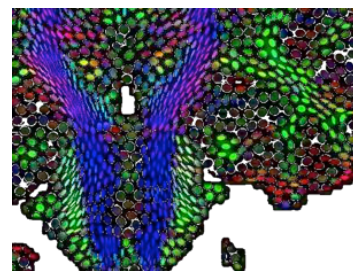
Textured slice

Images: Kratz, Breßler, Hotz, ZIB



3D Fiber tracking

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Glyph representations

Tensor field visualization – My view on the field

I Tensor as physical descriptors

Structure Tensor – Image Analysis

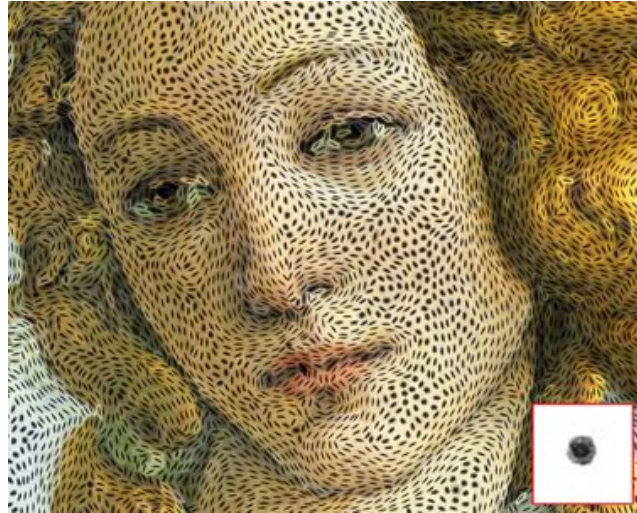


Image: Kratz, ZIB

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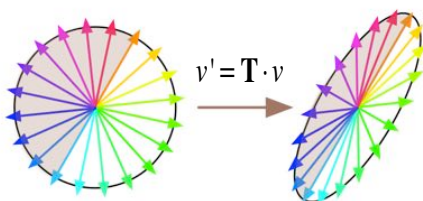
Tensor field visualization – My view on the field

I Tensor as physical descriptors

Different Characters of Tensors - Examples

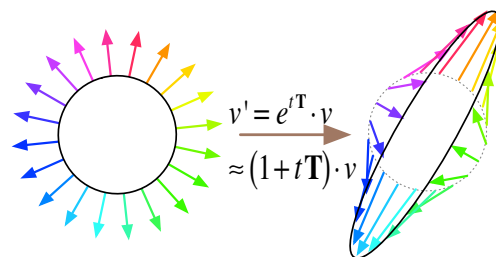
Describes a deformation

- Positive definite
- E.g. deformation tensor
- Determinant $\det \mathbf{T} \rightarrow$ volume change



Describes a generator of a deformation

- Indefinite
- E.g. stress tensor (forces/area)
- $\text{Trace} \mathbf{T} \rightarrow$ volume change



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I Tensor as physical descriptors

Tensor Invariants

- Play a fundamental role in the understanding of application specific tensors
 - Every application has its own invariants that are especially important. They often come with a domain specific language
- Relevant invariants should guide the choice of the visualization

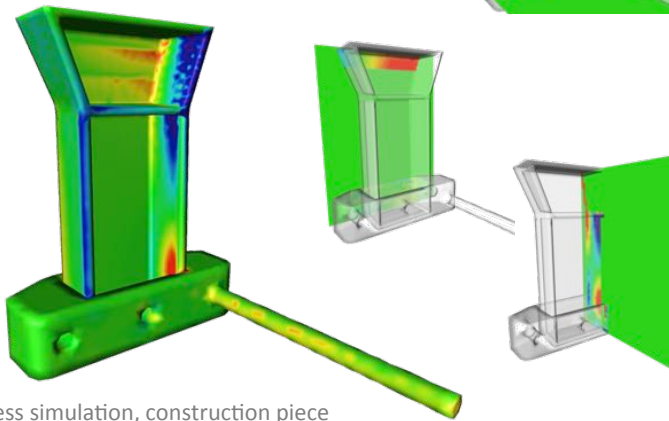
II Some basic visualization methods

What has visualization currently to offer

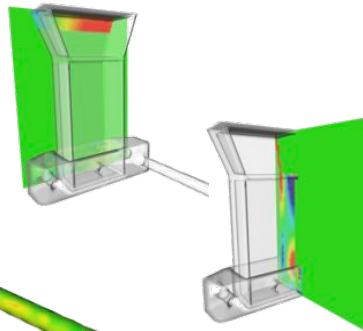
II Some basic visualization methods

Common practice

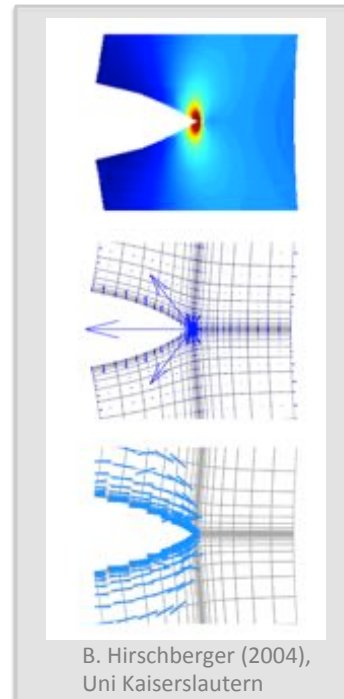
- Color Representation of derived scalars, e.g. trace
- 2D slices and surfaces



Stress simulation, construction piece
M. Stommel, Uni Saarbrücken



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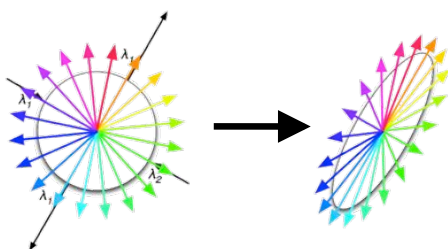


B. Hirschberger (2004),
Uni Kaiserslautern

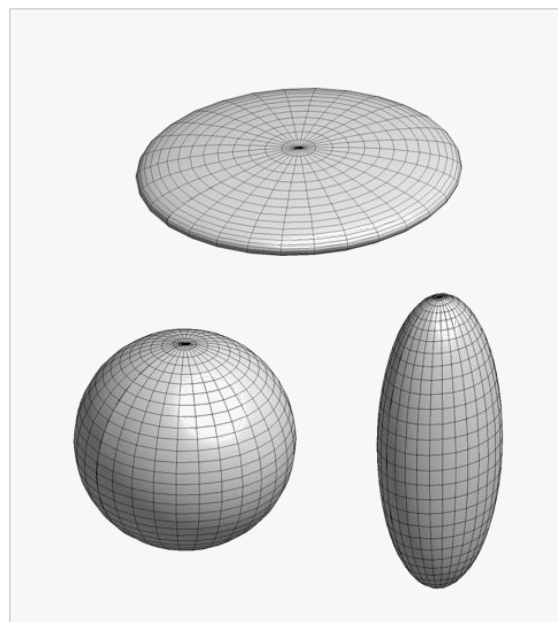
II Some basic visualization methods

Local Methods - Glyphs

- Geometric objects representing **tensor characteristics**
- Here: Ellipses
- Most frequently used visualization



The typical glyph: Ellipsoid

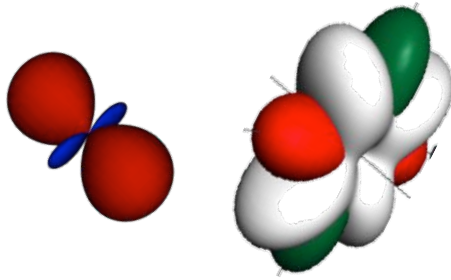


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II Some basic visualization methods

Local Methods - Glyphs

Many different glyph types in use



Example from Material Sciences

When to use?

- Evaluation of data quality
- Visualization overview
- Probing

Glyph design

- What should be represented?
- How should it be represented?

Glyph placement

- How many glyphs?
- Where to place them?

[Schultz2010, Kratz2014]

II Some basic visualization methods

Local Methods - Glyphs

How should it be represented – general design guidelines

- Preservation of symmetry
e.g. eigenvectors have no orientation, isotropic tensors have not distinguished direction
- Continuity
similar tensors should have similar glyph representations
- Disambiguate
tensors with different values should be reflected by different glyphs

[Schultz and Kindlmann 2010]

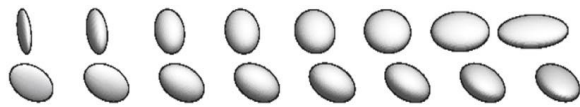
- Use glyphs that have been used in the community before

What should be represented – application specific guidelines

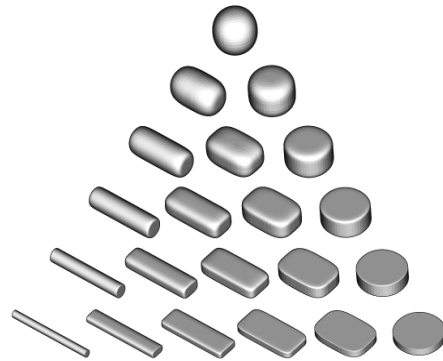
- Use application specific invariants for the design

II Some basic visualization methods

Local Methods - Glyphs



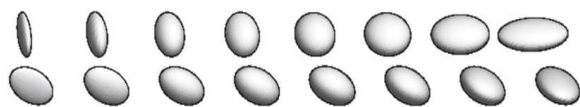
Resolving Perceptual Issues
Superquadrics



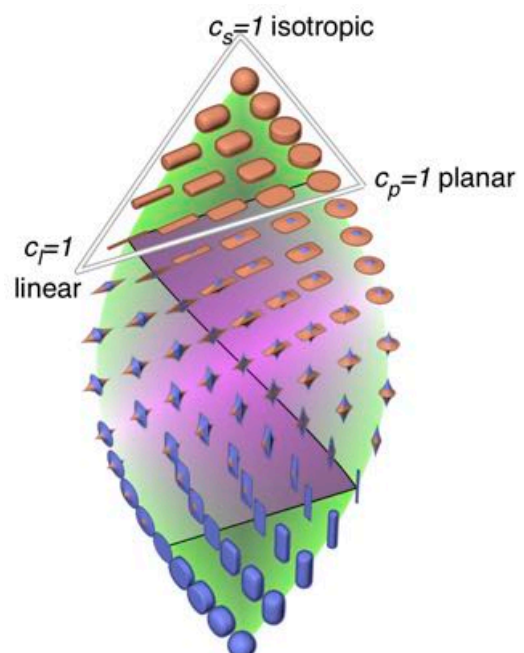
[Kindlmann2004]

II Some basic visualization methods

Local Methods - Glyphs



Resolving Perceptual Issues
Superquadrics



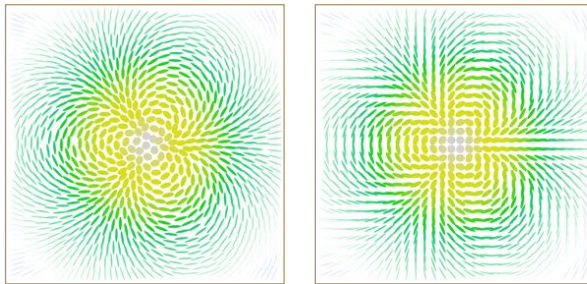
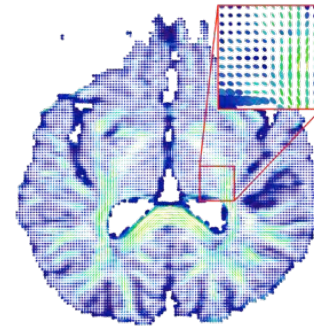
[Schulz2010]

II Some basic visualization methods

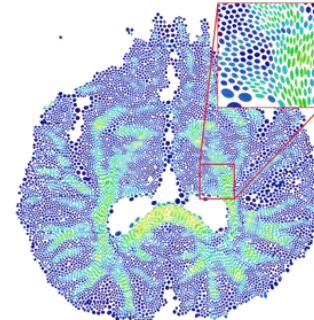
Local Methods - Glyphs

Glyph placement

- How many glyphs?
- Where to place them?



Anisotropic sampling



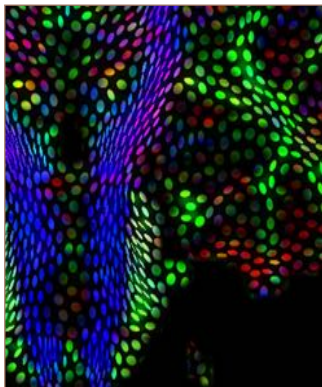
Images: Kratz, Bressler ZIB, Amira

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II Some basic visualization methods

Local Methods - Glyphs

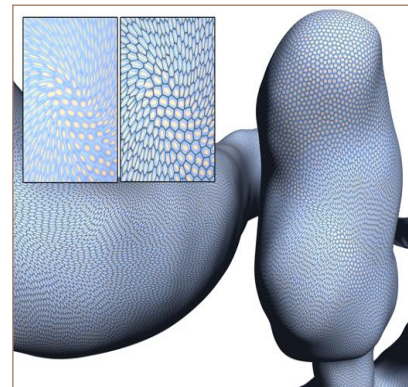
Particle based and geometric methods on planes, in 3d, on surfaces



[Feng2008]



[Kindlmann2006]



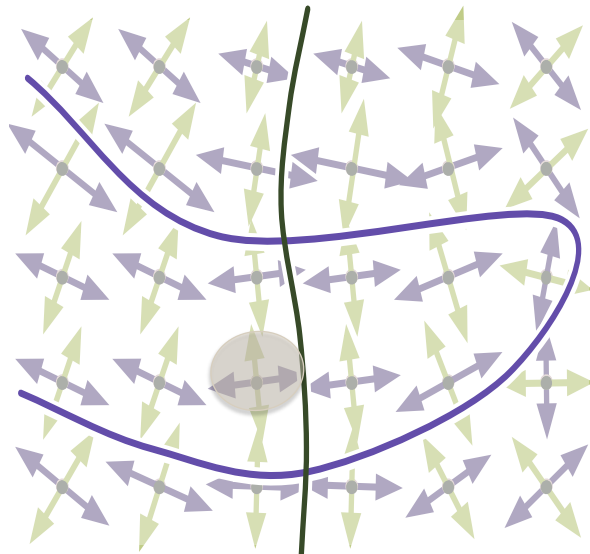
[Kratz2013]

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II Some basic visualization methods

Tensor lines

- Integral lines similar to streamlines
- Follow one eigenvector field



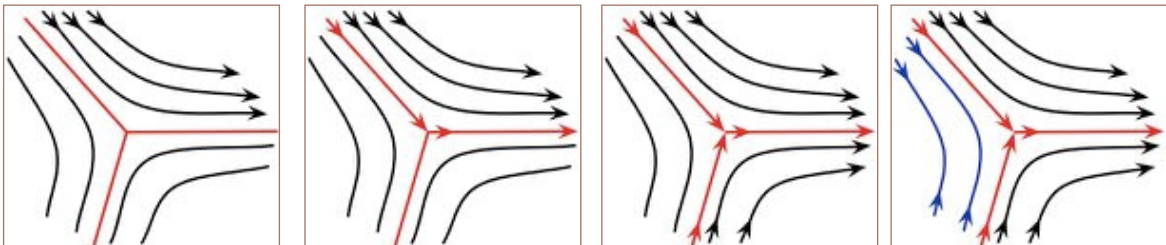
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II Some basic visualization methods

Tensor lines

Direction field is not orientable.
It is not a vector field!!

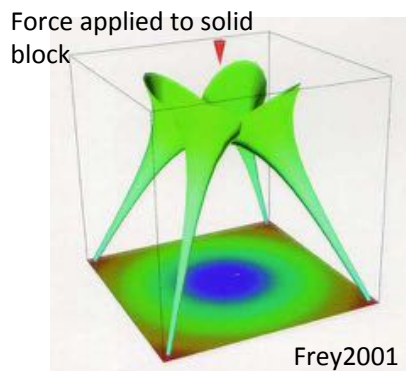


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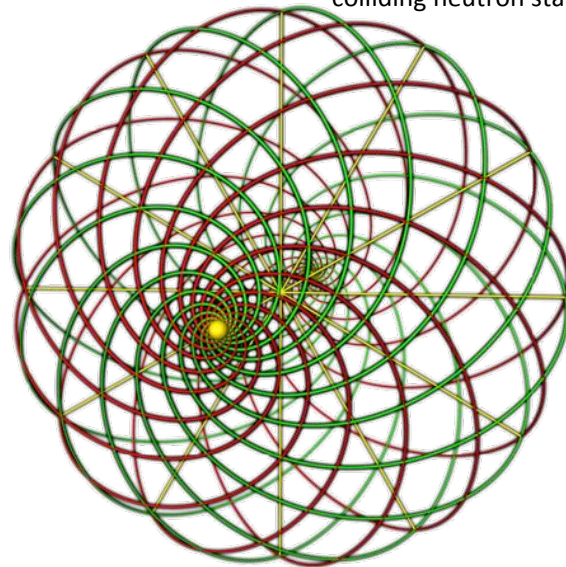
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II Some basic visualization methods

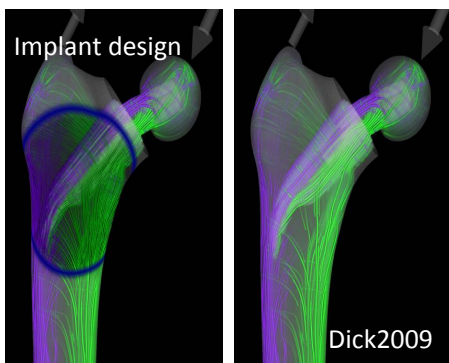
Tensor lines



Gravitational field of two colliding neutron stars



Images: Zobel, ZIB



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II Some basic visualization methods

Tensor lines – Fiber tracking

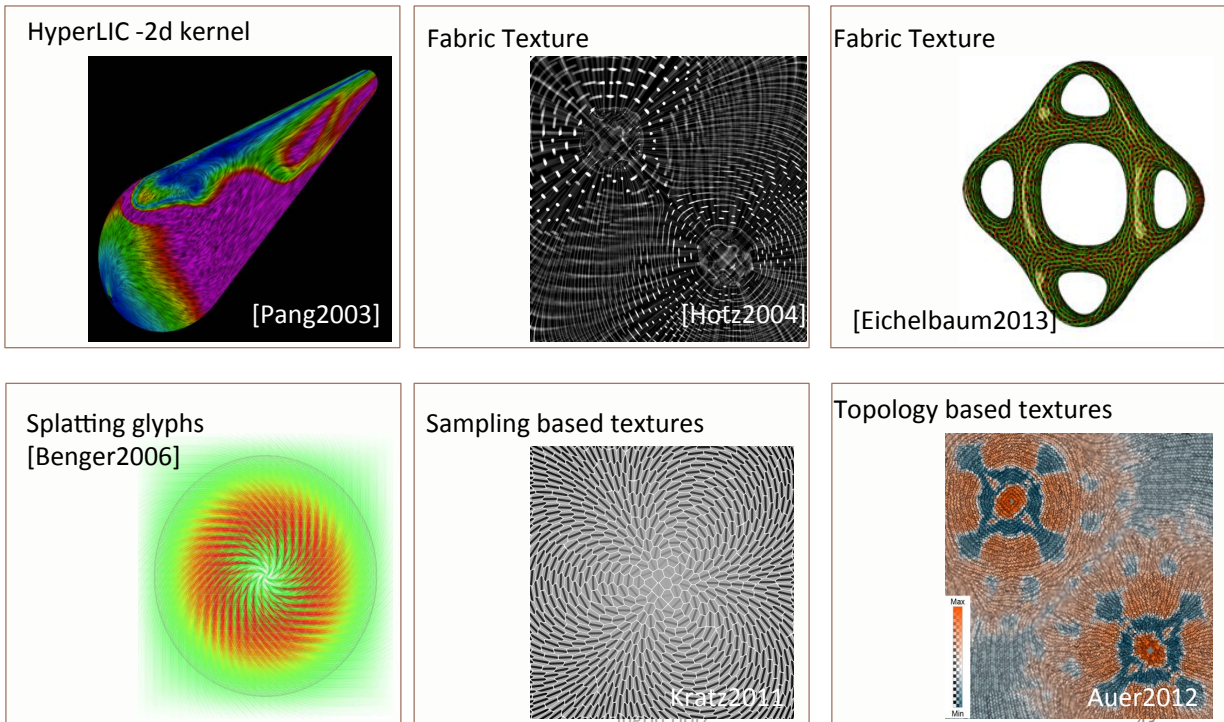
Tensors from diffusion MRI

Tracking of **neural fibers**

- Major tensor lines can be used to approximate fibers
- Line tracing only for regions of high anisotropy

II Some basic visualization methods

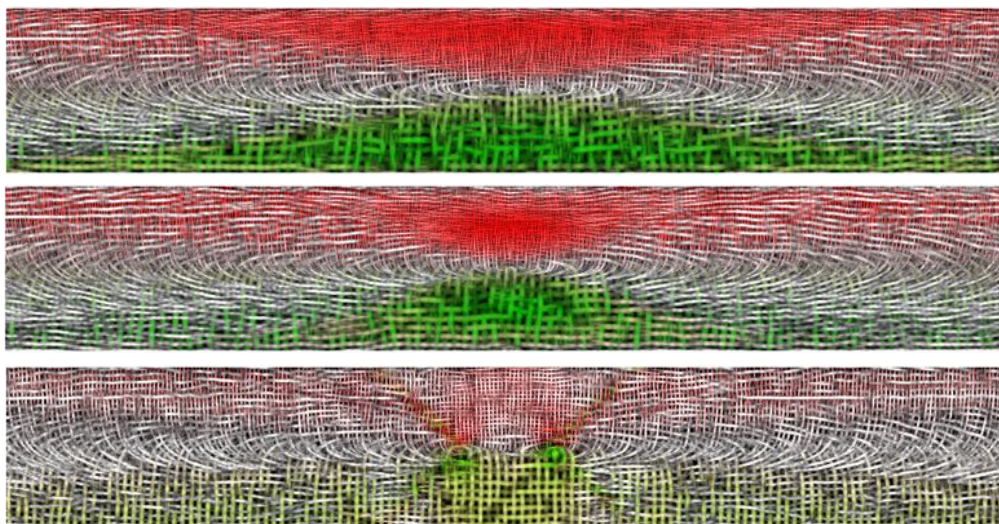
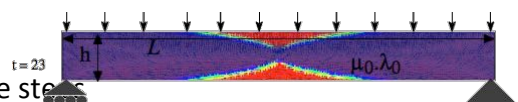
Textures



II Some basic visualization methods

Textures – Principal directions and values

- Beam bend under its own load.
- Simulation respecting evolving damage, 3 time steps
- Data: Louise Kellogg University of California.

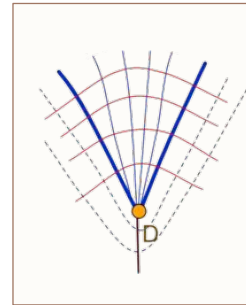
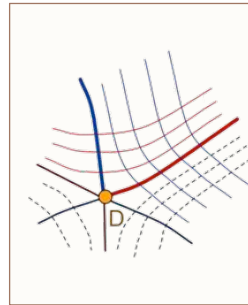
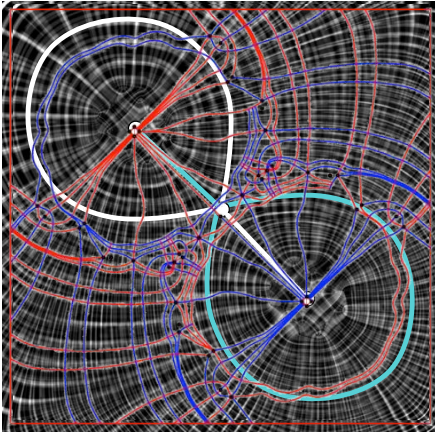


II Some basic visualization methods

Topology

Topological structure

- Segmentation of domain in areas of uniform directional behavior
- Similar to vector field topology – but different basic structures



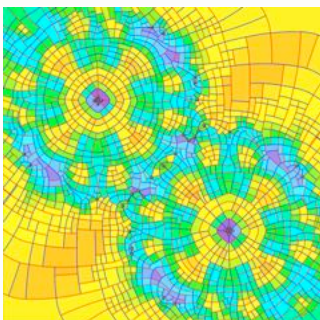
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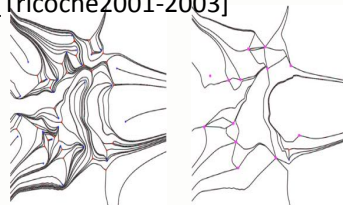
II Some basic visualization methods

Topology

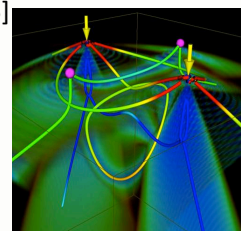
- Topology based Segmentation
- Topology based Textures [Auer2013]



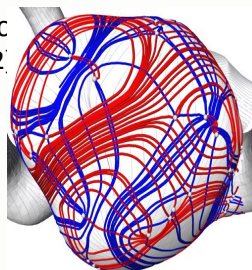
Computation, Simplification, Tracking [Tricoche2001-2003]



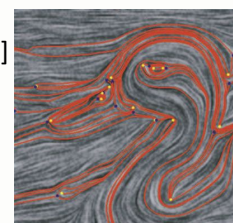
3D topology, critical lines [Zheng2005]



On Surface [Auer2012]



Tensor field design [Zhang2007]



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II Some basic visualization methods

What has been done so far?

Some technical stuff

- Glyphs from all perspectives
- **Technical generalization of vector field visualization methods**
- Textures

Much work on tensor processing

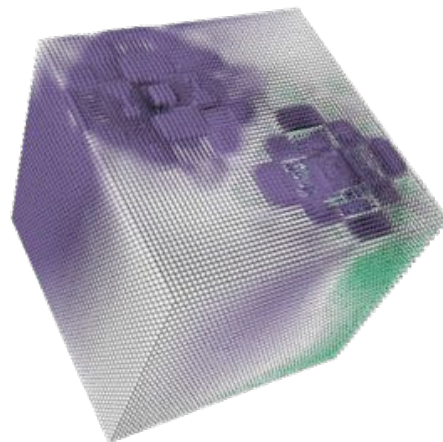
- Interpolation / reconstruction
- Morphology
- Topology

What is still missing

- Applications – link from visualization techniques to physical interpretations
- Notion of features
- Questions

III Tensor Invariants for feature definition

- **Exploarative framework**
supporting multiple applications
with different questions
- **Structuring the data**
 - Manage complexity of the data
 - Highlight trends
 - Point at critical/atypical behavior



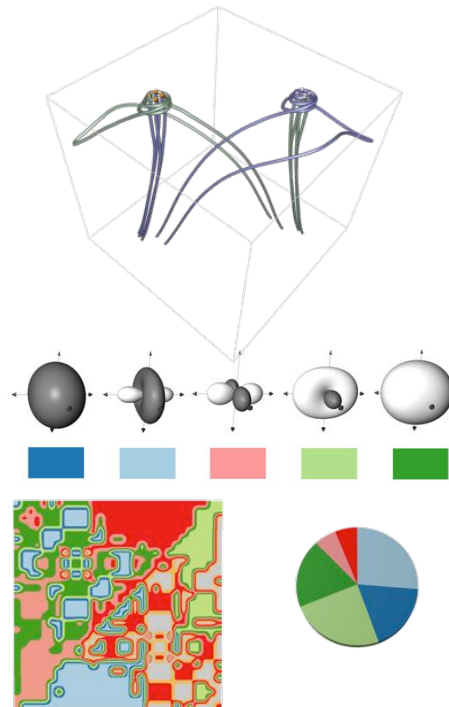
III Tensor Invariants for feature definition

- **Explorative framework**
supporting multiple applications
with different questions

→ **Domain specific feature spaces**

- **Structuring the data**
 - Manage complexity of the data
 - Highlight trends
 - Point at critical/atypical behavior

→ **Data atlas using a thumbnail like representation**



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III Tensor Invariants for feature definition

Tensor is uniquely defined by

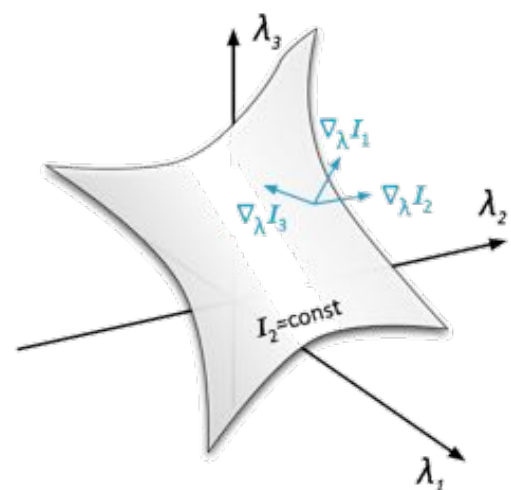
- 3 eigenvalues → point in 'shape-space'
- 3 eigenvectors → point in 'direction-space'

Tensor invariant

- Are the language of the applications

Use application specific invariants to parameterize the shape space

- **shape descriptors** $I_i(\lambda_1, \lambda_2, \lambda_3)$
- basis for the definition statistic views, glyph design, similarity measure

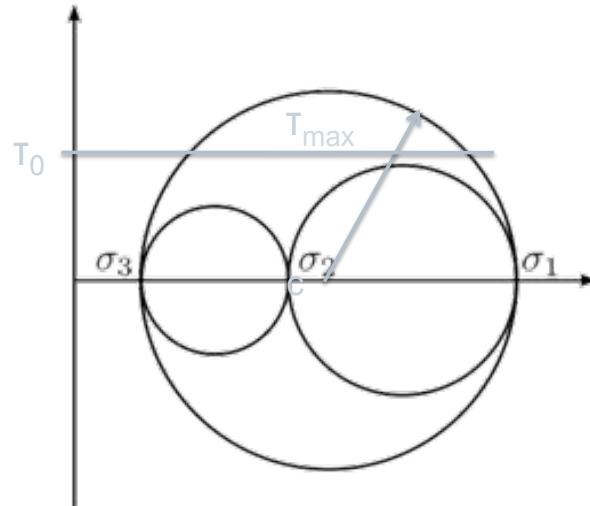


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The Stress Tensor and Failure Models

Example for failure analysis in mechanical engineering

→ Mohr Circle

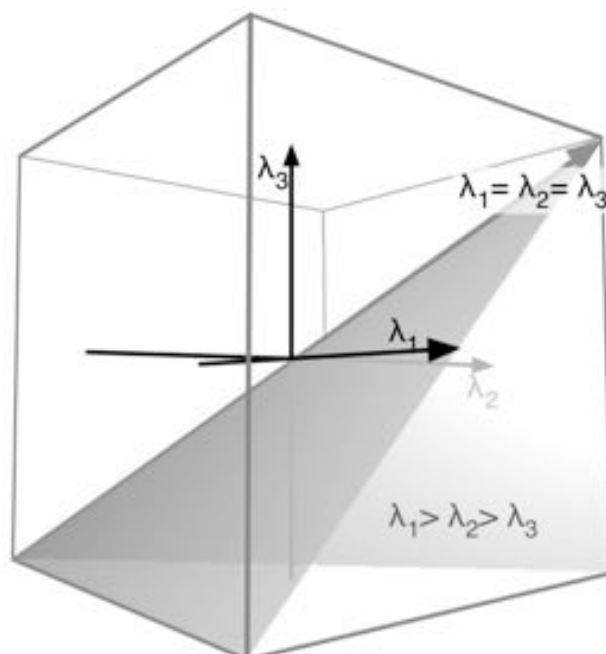


III Tensor Invariants for feature definition

Example

- Ordered shape space

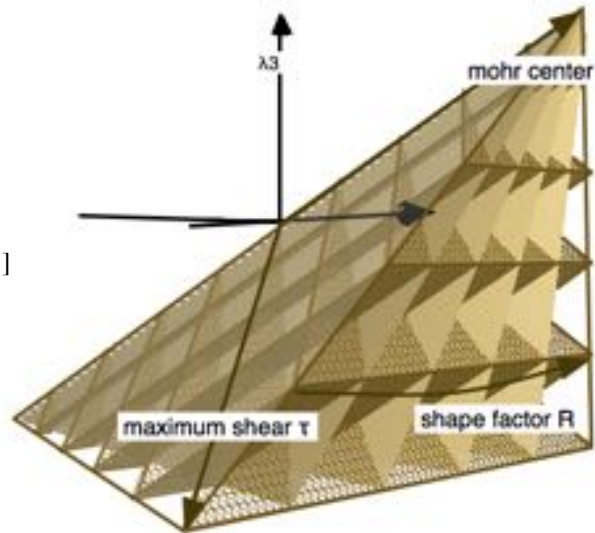
$$\lambda_1 \geq \lambda_2 \geq \lambda_3$$



III Tensor Invariants for feature definition

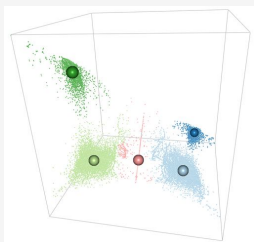
Example – Coulomb failure

- Parametrized by
 - Mohr center $c = \frac{\lambda_1 + \lambda_3}{2}$
 - Max Shear $\tau = \frac{\lambda_1 - \lambda_3}{2}$
 - Shape factor $R = \frac{\lambda_1 - \lambda_2}{\lambda_1 - \lambda_3} \in [0,1]$

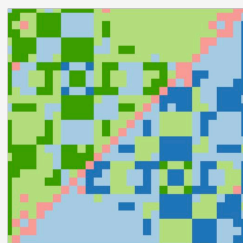


III Tensor Invariants for feature definition

Cluster in feature space

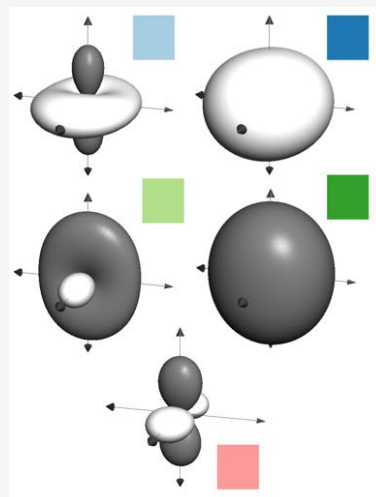


Spatial view: Top view of data



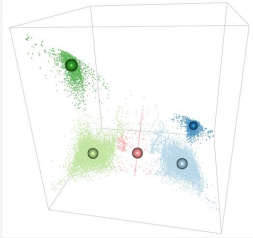
Cluster representatives

- Reynolds glyphs

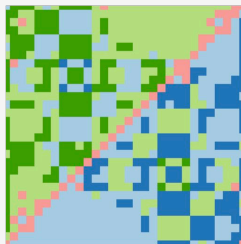


III Tensor Invariants for feature definition

Cluster in feature space

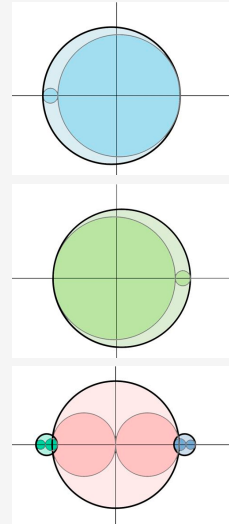


Spatial view: Top view of data



Cluster representatives

- Mohr circles

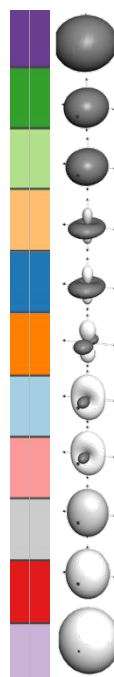
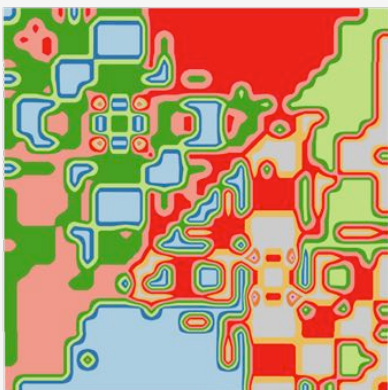


III Tensor Invariants for feature definition

Clustering

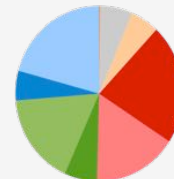
- Mean shift
- Full feature space + spatial coordinates

Top-view of the two-point load



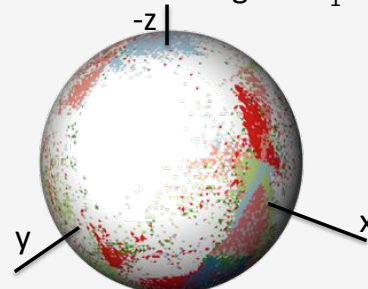
Cluster statistics

- Cluster size



■+ ■ < 1%

- Directional histogram λ_1

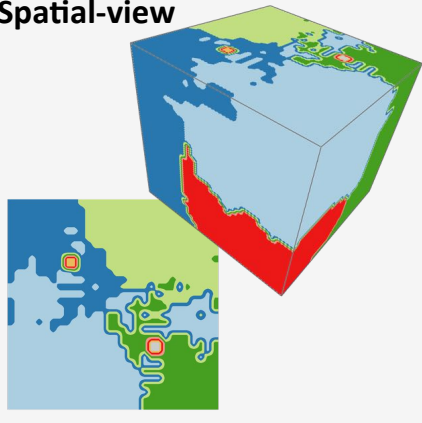


III Tensor Invariants for feature definition

Clustering

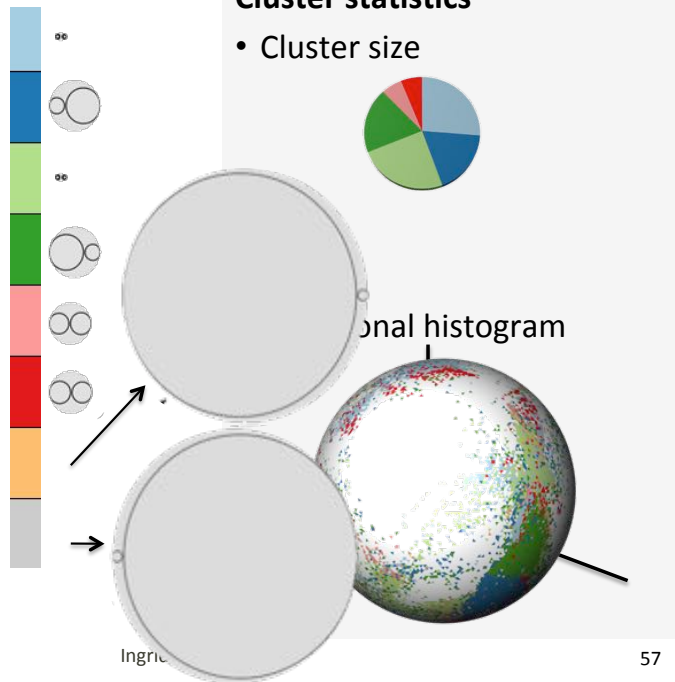
- Mean shift
- Feature space (shear, shape) + spatial coordinates

Two-point load Spatial-view

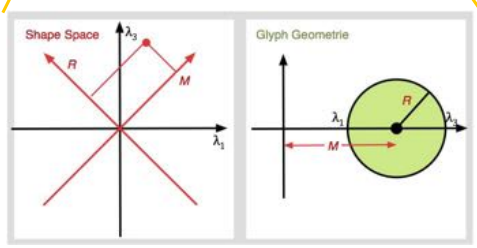
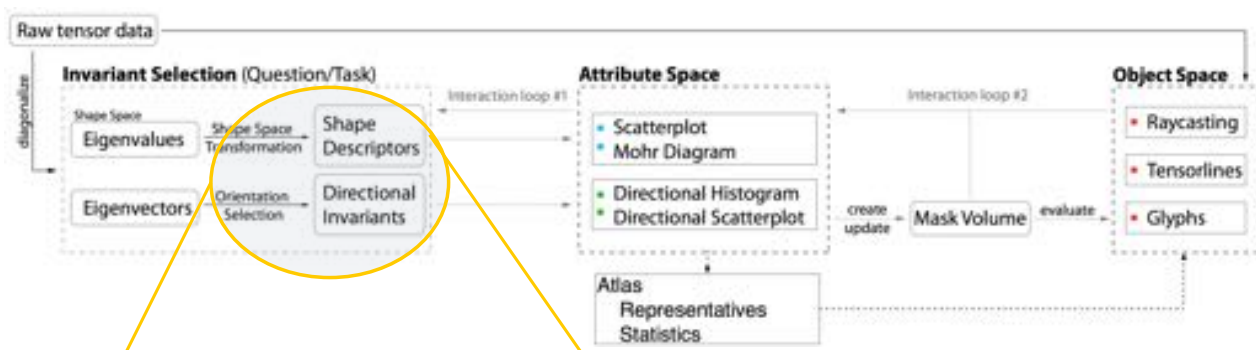


Cluster statistics

- Cluster size



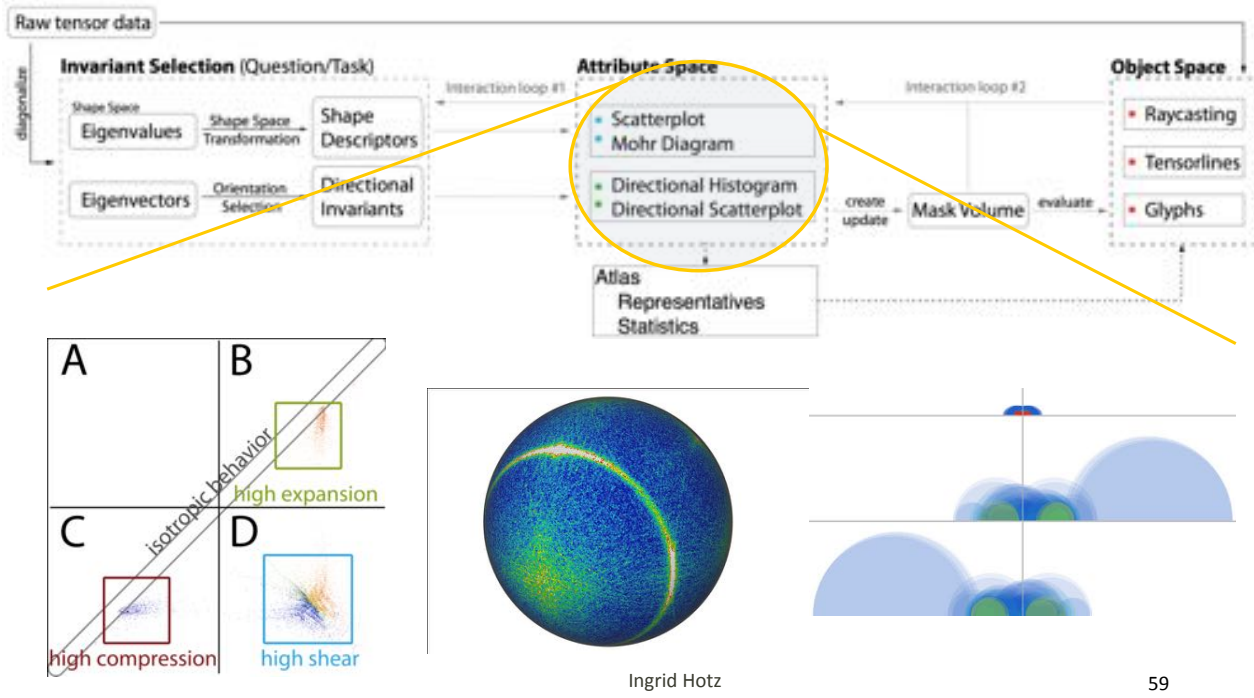
III Tensor Invariants for feature definition



Feature Space Selection

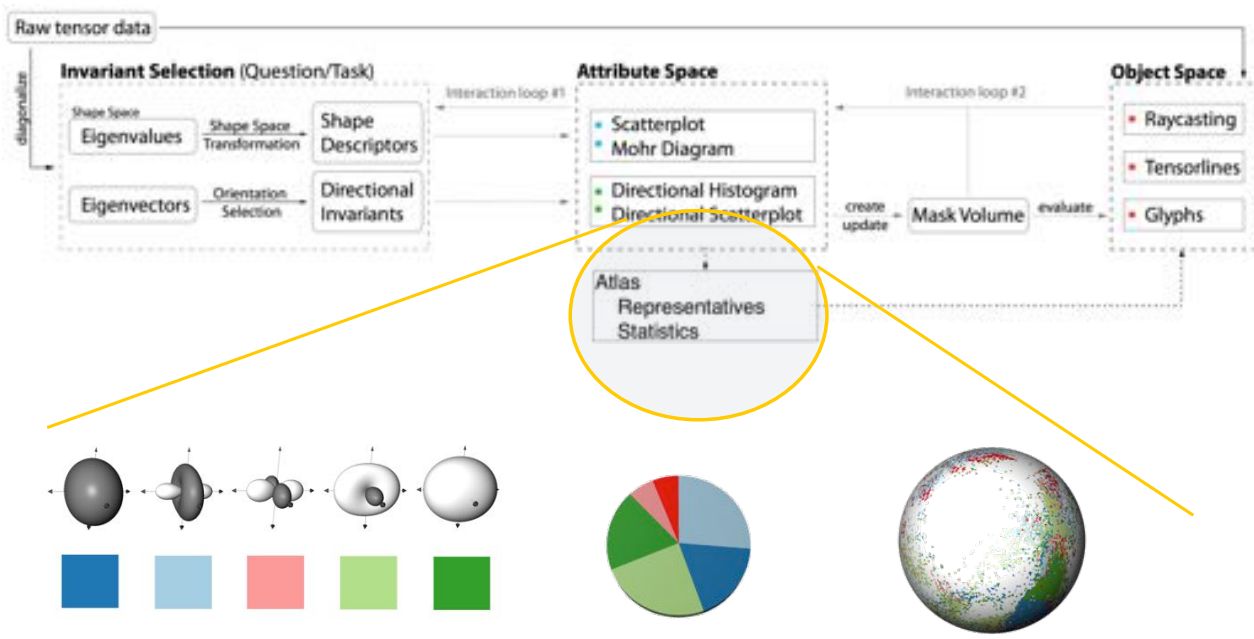
- Change of basis results in another set of shape descriptors
- Determines
 - Exploration space
 - Metric, similarity measure
 - Glyph design

III Tensor Invariants for feature definition



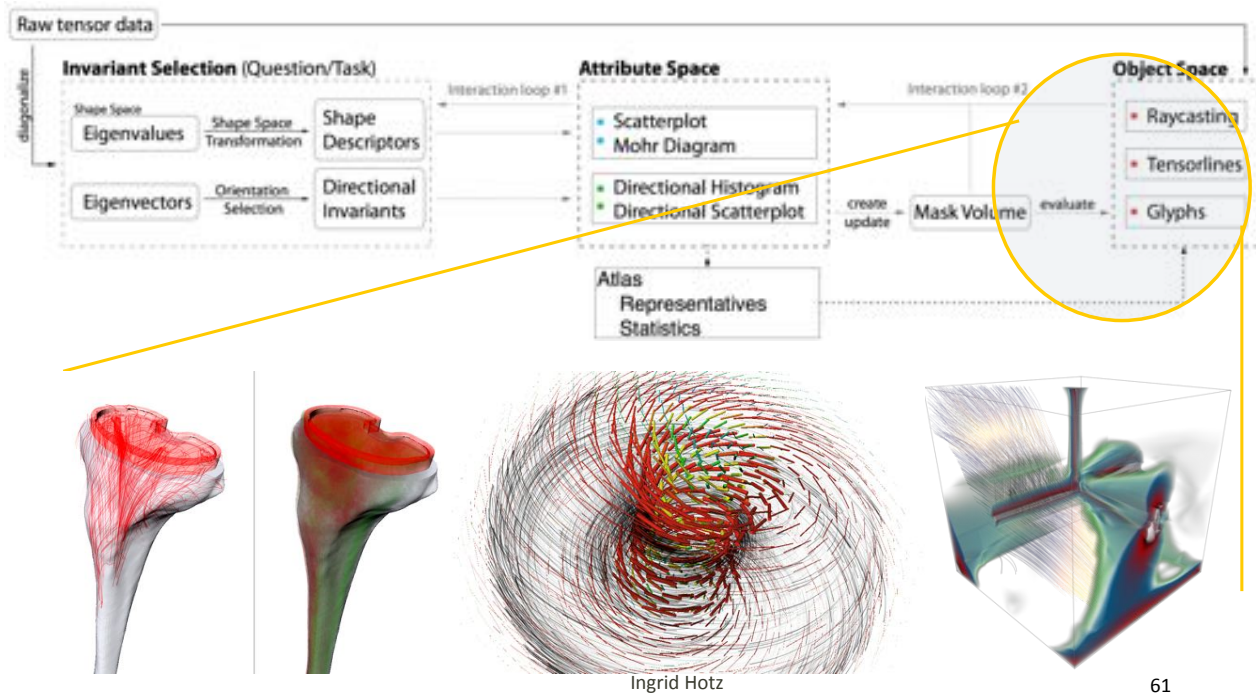
Ingrid Hotz

III Tensor Invariants for feature definition



Ingrid Hotz

III Tensor Invariants for feature definition



Ingrid Hotz



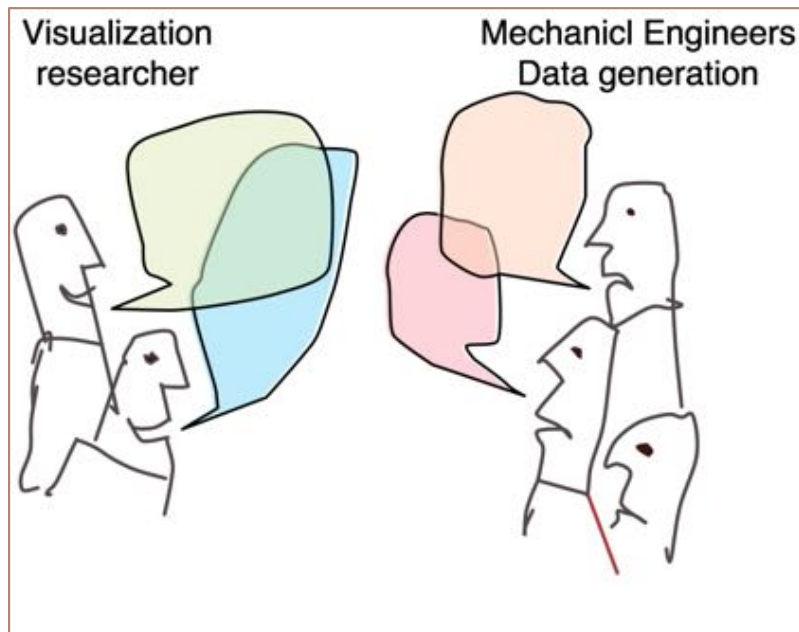
Tensor Visualization Driven
Mechanical Component Design

or
The story of a collaboration

IV Story of a Collaboration

Starting point

- Unspecific goals
- Different language



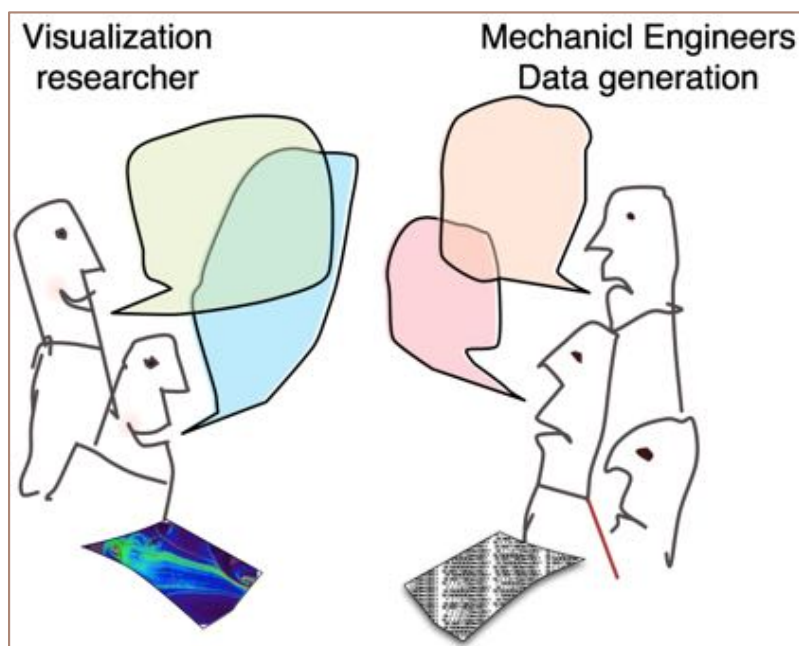
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IV Story of a Collaboration

First Experiments

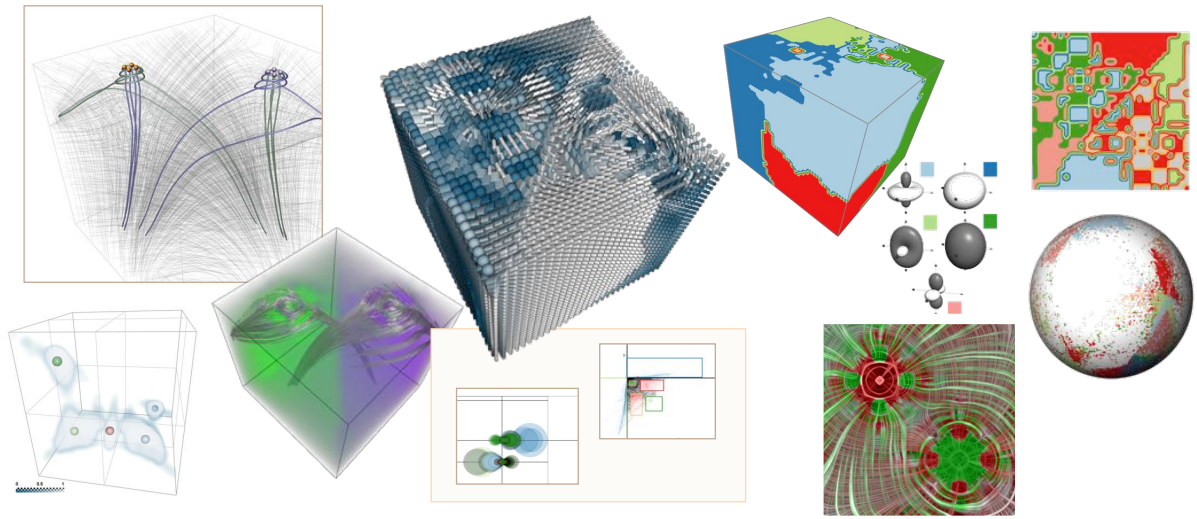
- Everybody shows what they have



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- Visualization – Framework



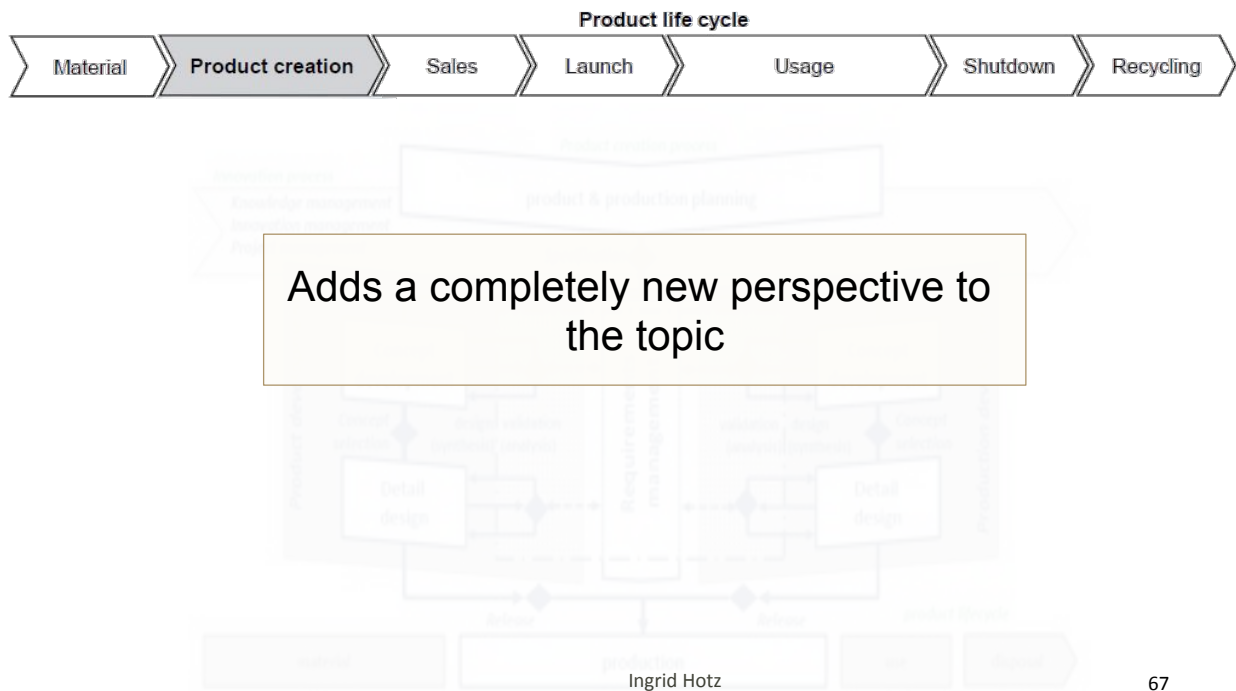
Images: Kratz, ZIB, Amira

- Engineering – Product Development Process



Tensor field visualization – My view on the field

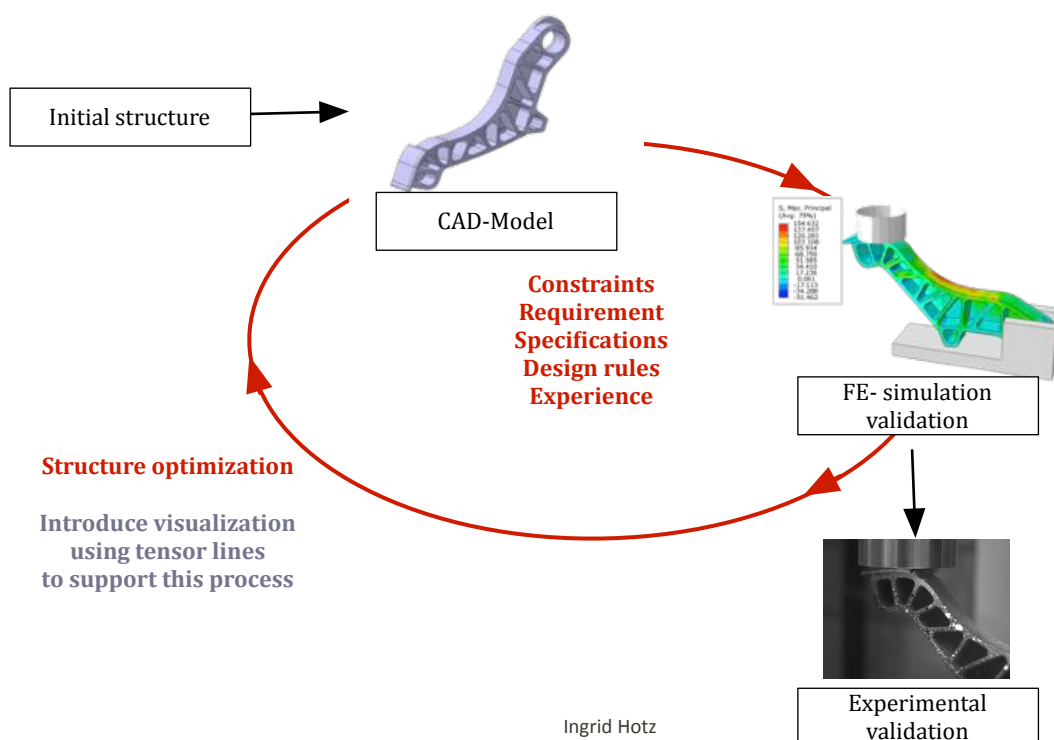
Tensors in Material Sciences Engineering – Product Development Process



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Tensor field visualization – My view on the field

Engineering – Product Development Process



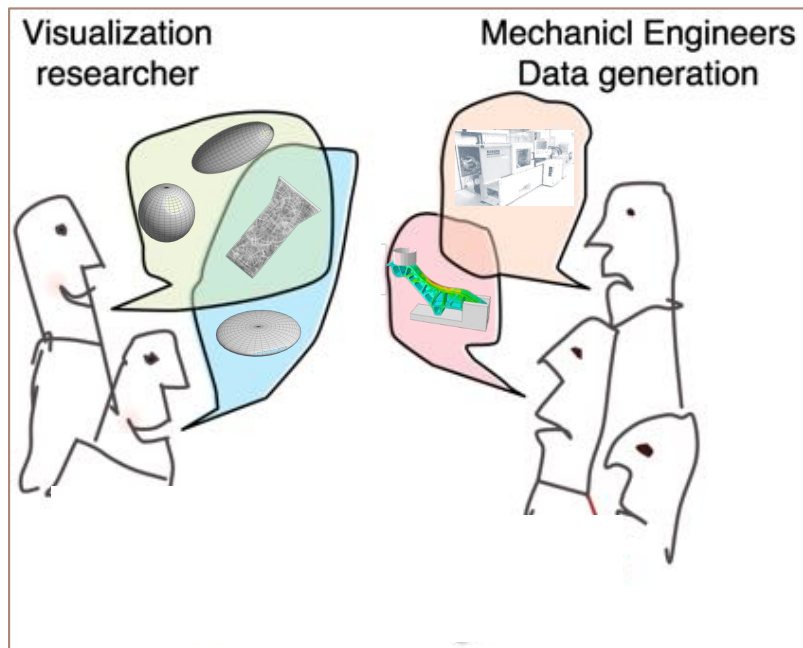
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IV Story of a Collaboration

First Experiments

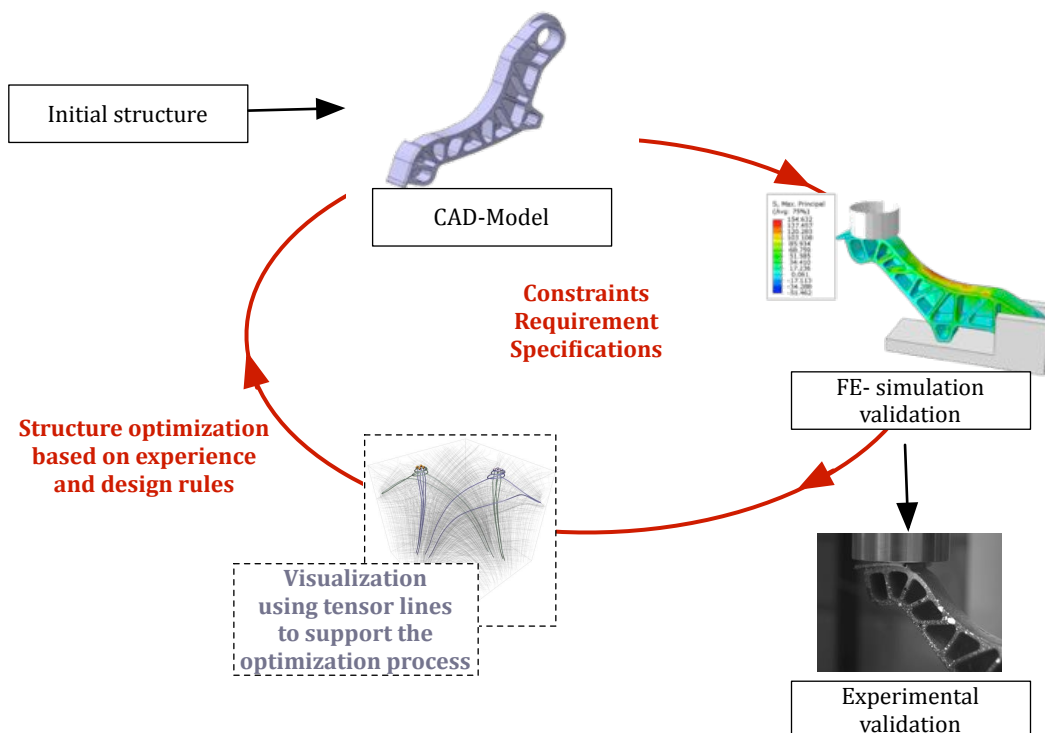
- Everybody shows what they have



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IV Story of a Collaboration



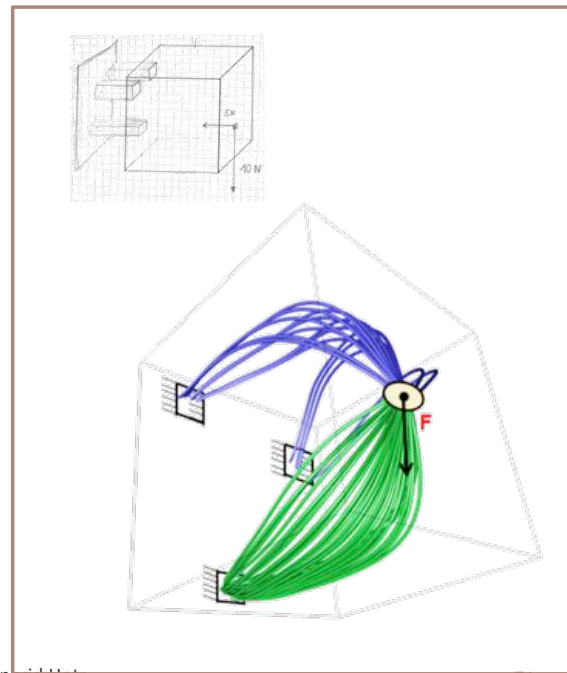
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IV Story of a Collaboration

Hypothesis resulting from first experiments and discussions

- The **tensor lines** for a stress tensor field are related to the **major load paths** from the **operating loads to the fixation** points of a technical part
- Tensor lines can be used to support the **design of reinforcement structures**



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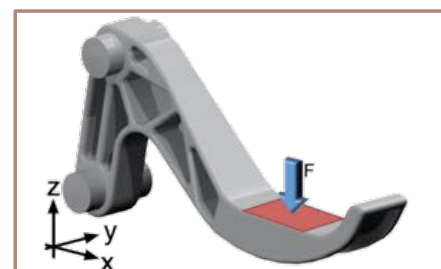
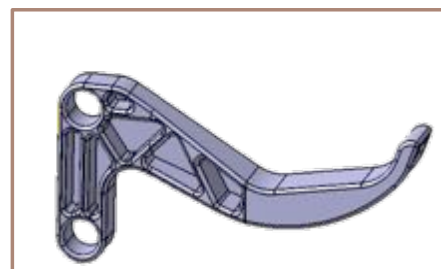
IV Story of a Collaboration

Design of a break lever

- Build reinforcement structure on basis of tensor line visualization
- **Combination of an automatic analysis with the expert's knowledge**
- Manageable but realistic case

Evaluation

- Comparison to reference structure – same volume
- Numerical validation
- Experimental validation



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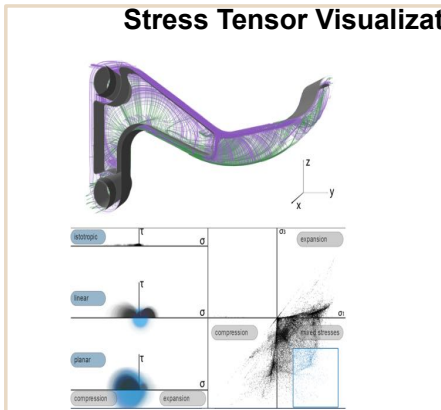
72

IV Story of a Collaboration

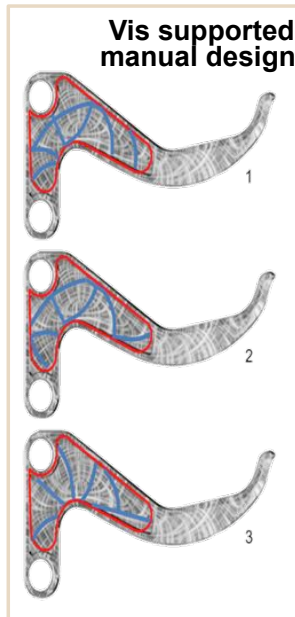
Design of the rib structure

- New geometries based on tensor visualization
- No additional steps in the workflow

Stress Tensor Visualization



Vis supported manual design



CAD - Model



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IV Story of a Collaboration

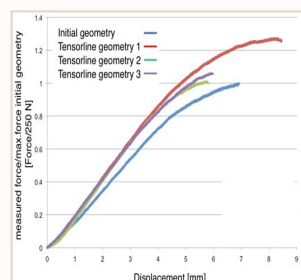
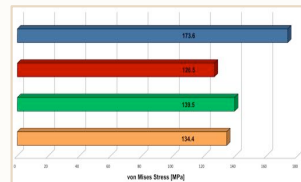
Numerical and experimental tests

- New designs (red, green, yellow)
- Reference geometry (blue)

→ All new geometries performed better as the reference geometry



Rapid Prototyping and experimental validation: Test setup



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IV Story of a Successful Collaboration

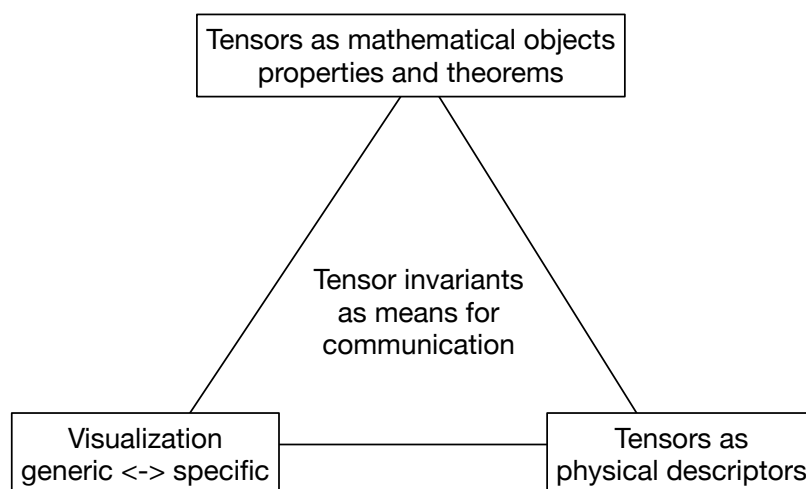
Advances in both fields

Continuing Story

- Many exiting questions and ideas
- Much fun
- Proposal submitted



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Relevance and meaning of mathematical properties and theorems is application dependent

- *Anisotropic Sampling of Planar and Two-Manifold Domains for Texture Generation and Glyph Distribution*, Kratz, Baum, Hotz, TVCG, 2013
- *Three-Dimensional Second-Order Tensor Fields: Exploratory Visualization and Anisotropic Sampling* (phdthesis), Andrea Kratz, 2013
- *Visualization and Analysis of Second-Order Tensors: Moving Beyond the Symmetric Positive-Definite Case*, Kratz, Auer, Stommel, Hotz, Computer Graphics Forum - State of the Art Reports, 2013
- *Tensor Invariants and Glyph Design*, Kratz, Auer, Hotz, Visualization and Processing of Tensors and Higher Order Descriptors for Multi-Valued Data (Dagstuhl'11), Springer, 2014
- **<http://www.tensorvis.org/>**