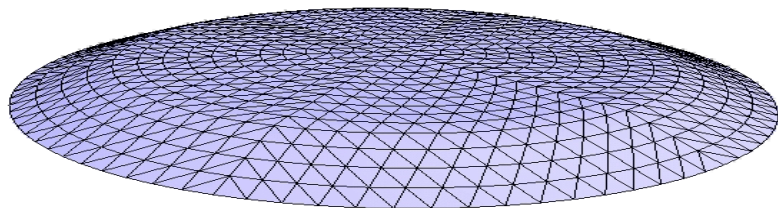




Spectral Shape Analysis with Applications in Medical Imaging



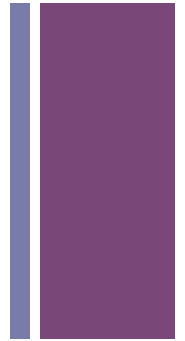
SIAM Annual Meeting 2013
San Diego

Martin Reuter – reuter@mit.edu

Mass. General Hospital, Harvard Medical, MIT

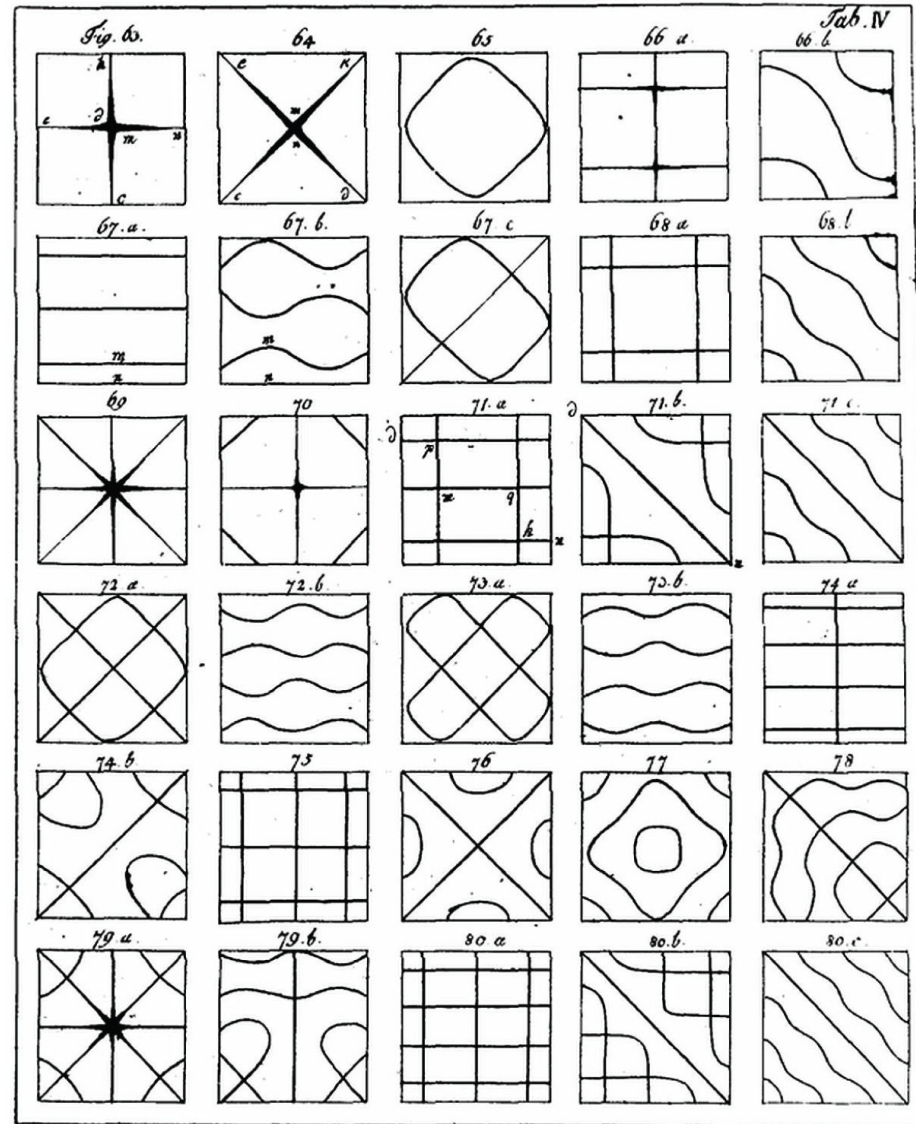


+ Sound and Shape



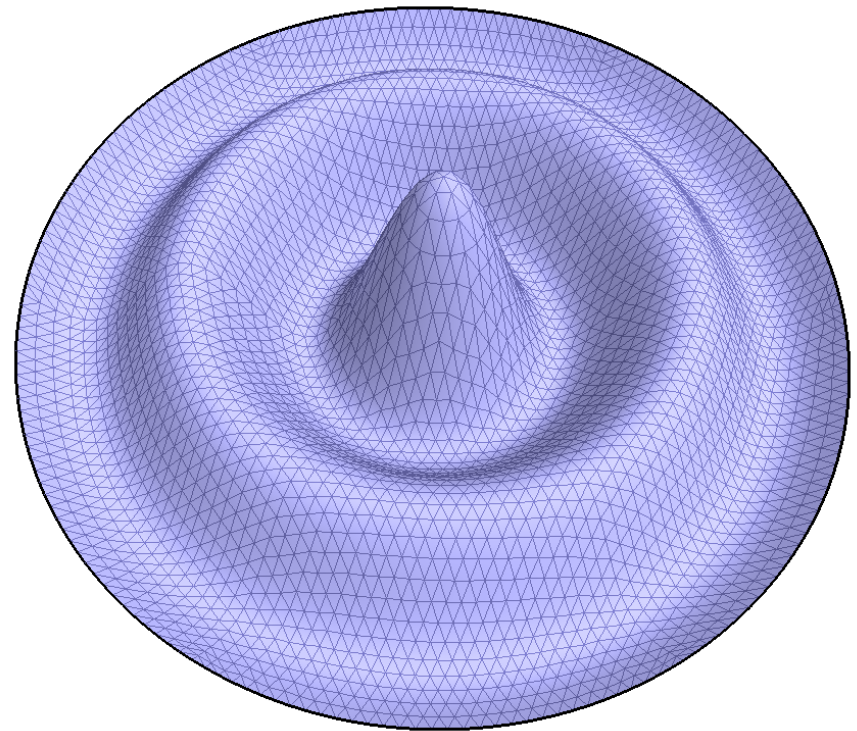
+ Chladni's strange patterns

- vibration of plates
- used a violin bow
- discovered sound patterns by spreading sand on the plates
- 1809 invited by Napoleon
- who held out price for mathematical explanation
- “Entdeckungen über die Theorie des Klanges” (Discoveries concerning the theory of music), Chladni, 1787



+ Can one “hear” Shape?

- First asked by Bers, then paper by Kac 1966, idea dates back to Weyl 1911 (at least).
- The sound (eigen-frequencies) of a drum depend on its shape.
- This spectrum can be numerically computed if the shape is known.
- E.g., no other shape has the same spectrum as a disk.
- Can the shape be computed from the spectrum?



+ Laplace Spectrum

Definition

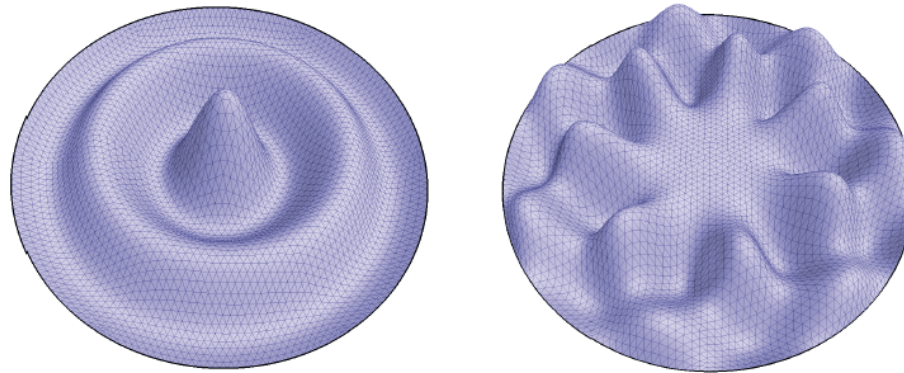
Helmholtz Equation (Laplacian Eigenvalue Problem):

$$\Delta f = -\lambda f, \quad f : M \rightarrow \mathbb{R}$$

Solution: Eigenfunctions f_i with corresponding family of eigenvalues (**Spectrum**):

$$0 \leq \lambda_1 \leq \lambda_2 \leq \cdots \uparrow +\infty$$

Here Laplace-Beltrami Operator: $\Delta f := \operatorname{div}(\operatorname{grad} f)$

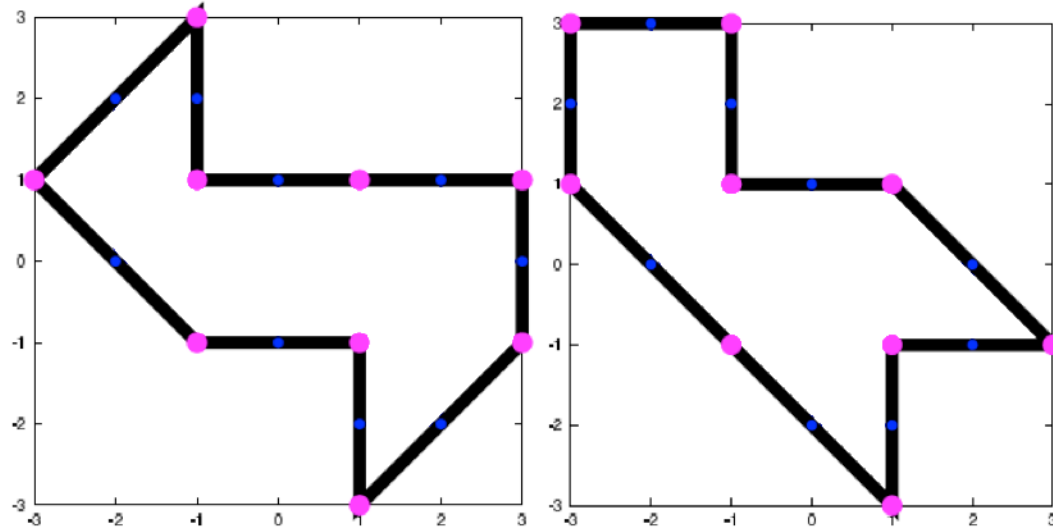


+ Can one hear Shape?

Answer

No! Isospectral drums exist (Gordon, Webb, Wolpert - 1992)

- rare
- concave in 2D
- only pairs

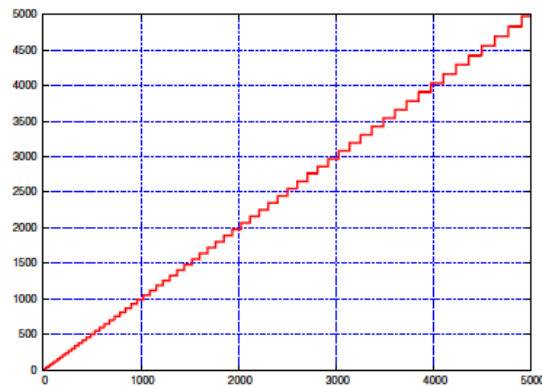


Geometry

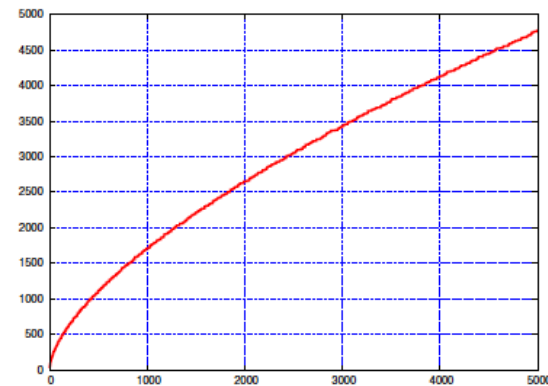
Nevertheless, they share area, boundary length, genus...

+ Weyl's Theorem

EW 2D-Sphere



EW 3D-Cube

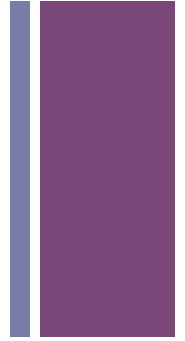


Theorem (Weyl - 1911,1912)

$$\lambda_n \sim \frac{4\pi}{\text{area}(D)} n \quad \text{for } d = 2 \text{ and } n \rightarrow \infty$$

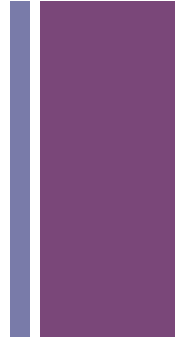
$$\lambda_n \sim \left(\frac{6\pi^2}{\text{vol}(D)} \right)^{\frac{2}{3}} n^{\frac{2}{3}} \quad \text{for } d = 3 \text{ and } n \rightarrow \infty.$$

+ Heat Trace Expansion



- More Geometric and Topological Information:
- Riemannian Volume
- Riemannian Volume of the Boundary
- Euler Characteristic for closed 2D Manifolds
- Number of holes for planar domains
- Possible to extract data numerically from beginning sequence [reuter:06] (500 Eigenvalues)

+ Contributions



Shape Analysis: Reuter, Wolter, Peinecke [SPM05], [JCAD06] (most cited paper award 09) and Patent Appl.

- Introduced Laplace-Beltrami Op. for Non-Rigid Shape Analysis.
- Cubic FEM to obtain accurate solutions for surfaces and solids.
- Before a mesh Laplace (simplified linear FEM) has been used for parametrization, smoothing, mesh compression

Image Recognition: Peinecke et al [JCAD07] (Mass Density LBO)

Neuroscience Applications:

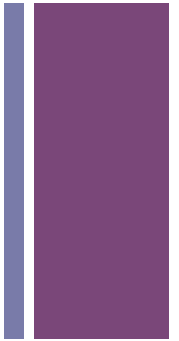
- Statistical morphometric studies of brain structures (eigenvalues), Niethammer, Reuter, Shenton, ... [MICCAI07], Reuter.. [CW08]
- Topological studies of eigenfunctions, Reuter.. [CAD09]

Segmentation: Reuter.. (IMATI, Genova) [SMI09] [IJCV09]

Correspondence: Reuter [IJCV09]

+ Finite Element Method

- Instead of graph/wireframe (vertices, edges), we look at elements that assemble our geometry without gaps:
 - triangles
 - tetrahedra
 - voxels....
- We define basis functions over this discretized geometry (linear, quadratic, cubic ...)
- We get a powerful framework to solve differential equations (not just Laplace).
- We also developed cubic FEM for parametrized surfaces (working in parameter space, similar to isoparametric FEM) [Reuter CAD06]



+ Higher Order

Theorem (Convergence)

*For decreasing mesh size h and order p form functions:
Eigenvalues converge with order*

$$O(2p)$$

and Eigenfunctions with order

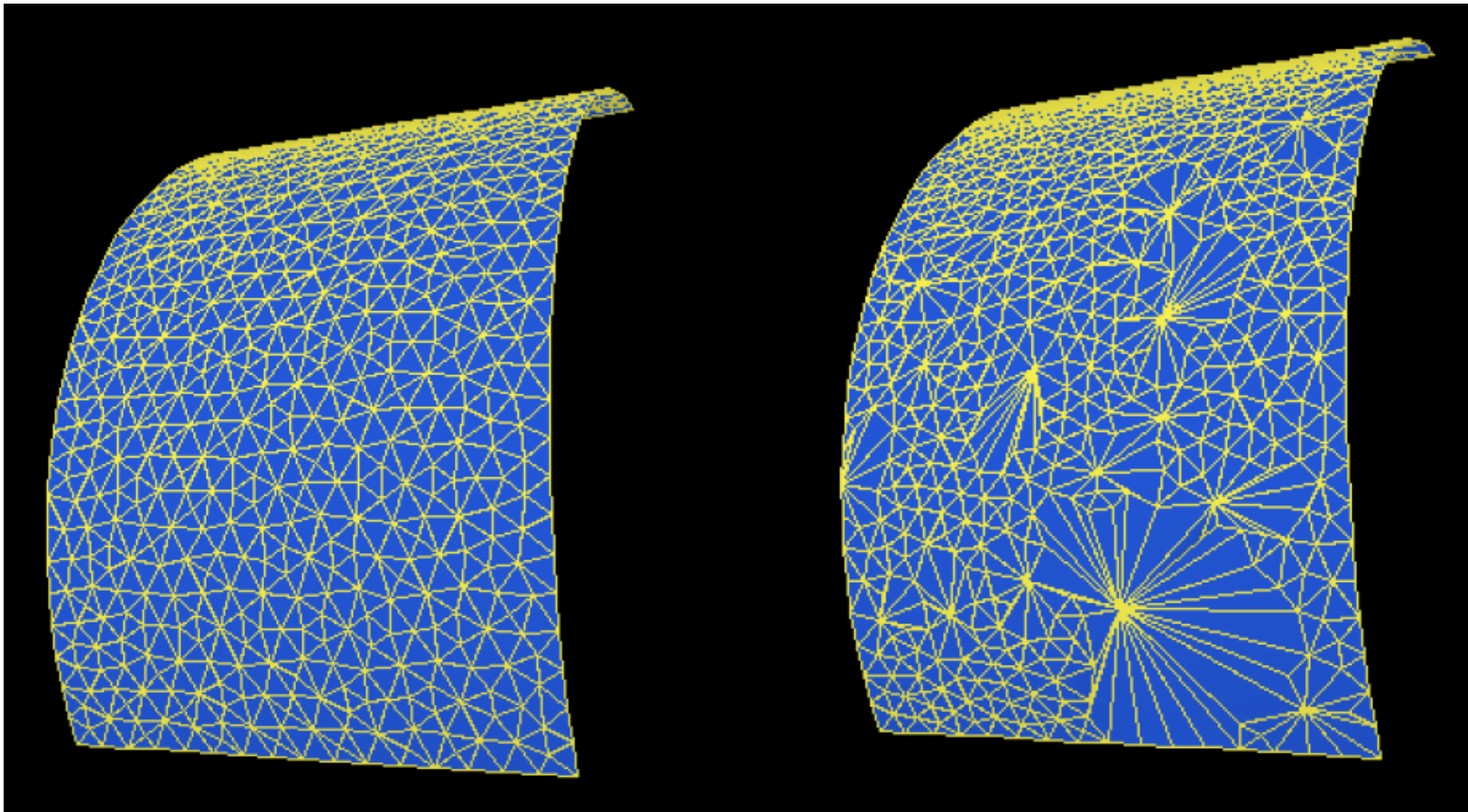
$$O(p + 1)$$

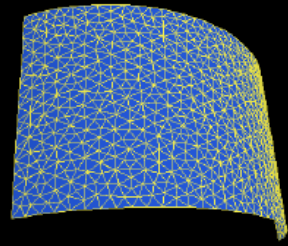
in the L_2 norm .

⇒ Always prefer higher order FEM approximations over mesh refinement.

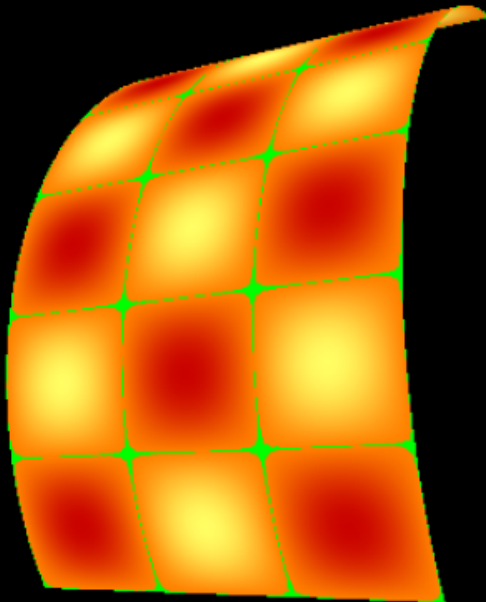
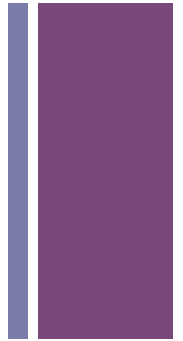
+ Uniform and Non-Uniform Mesh

[SMI09]

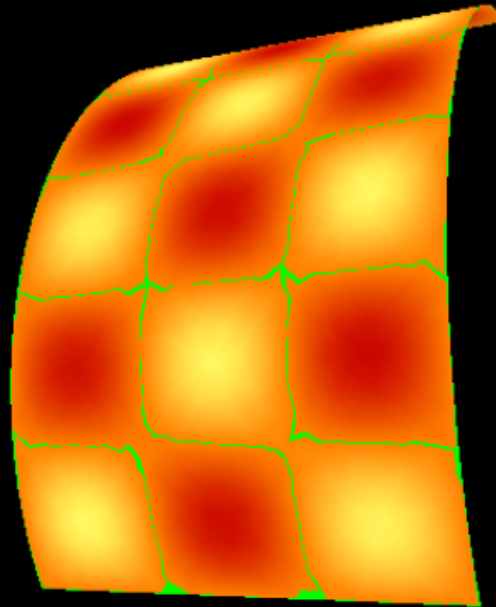




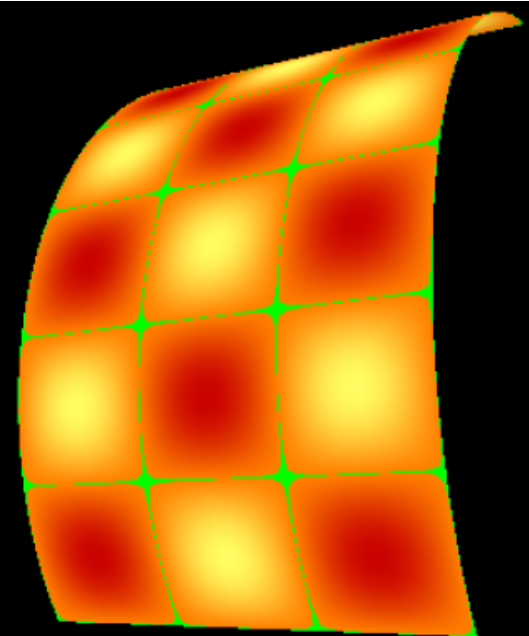
Uniform Mesh (Efunc 23):



real

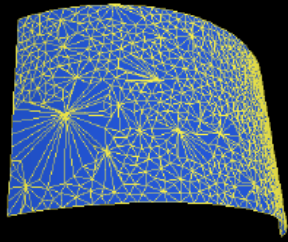


linear

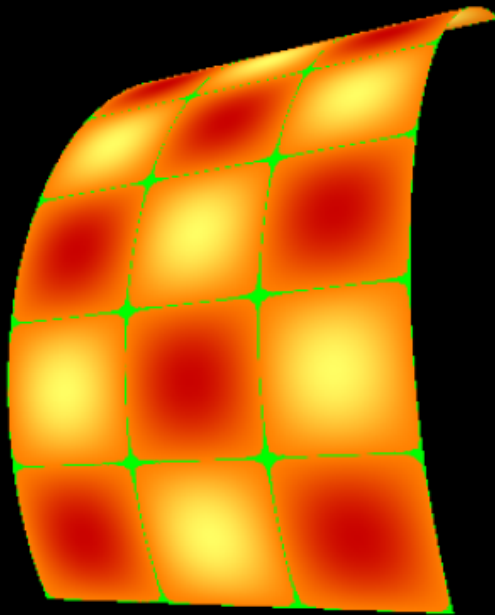
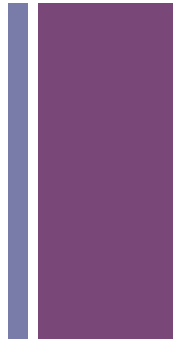


cubic

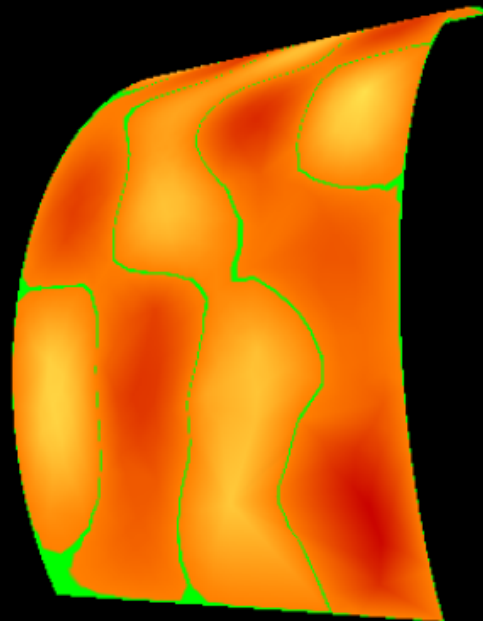
[SMI09]



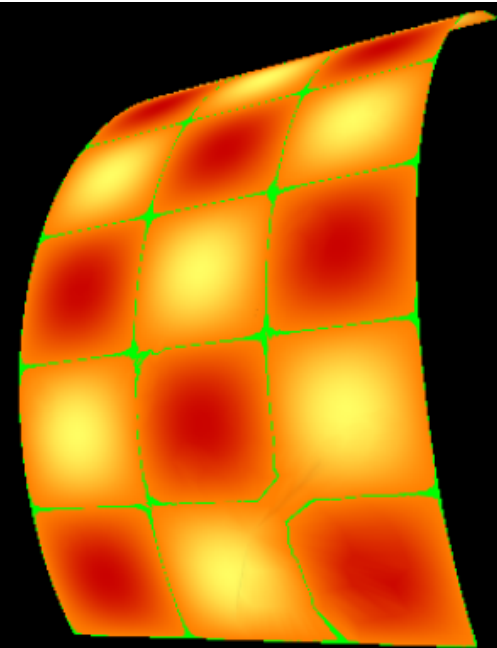
Non-Uniform Mesh (Efunc 23):



real

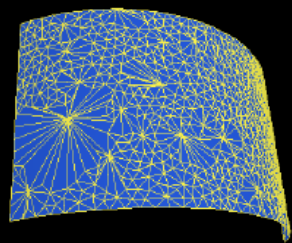


linear

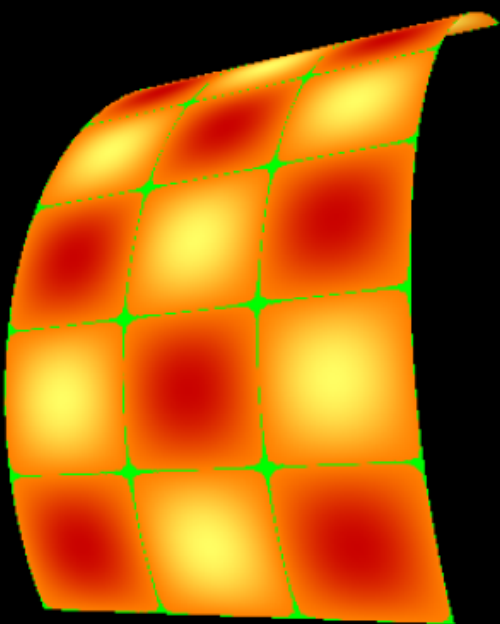
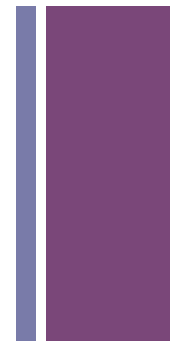


cubic

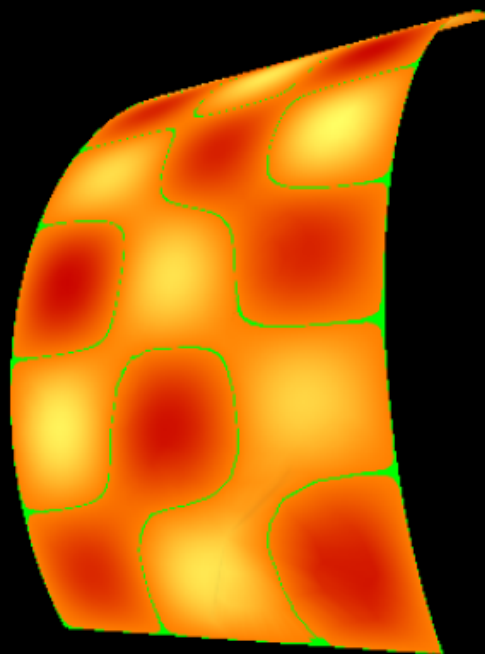
[SMI09]



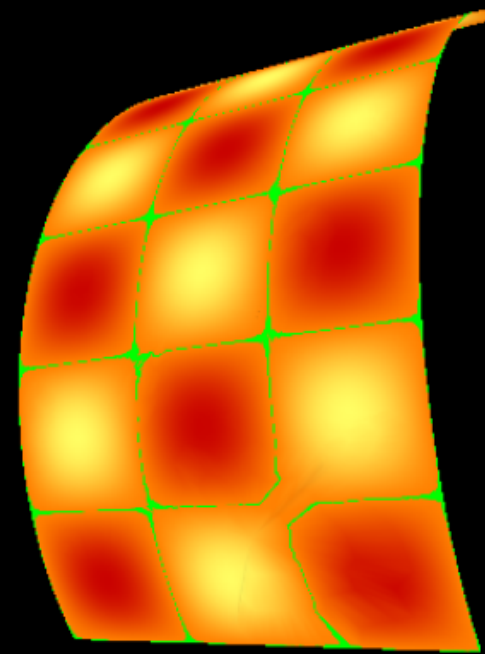
Non-Uniform Mesh same DOF (Efunc 23):



real



linear

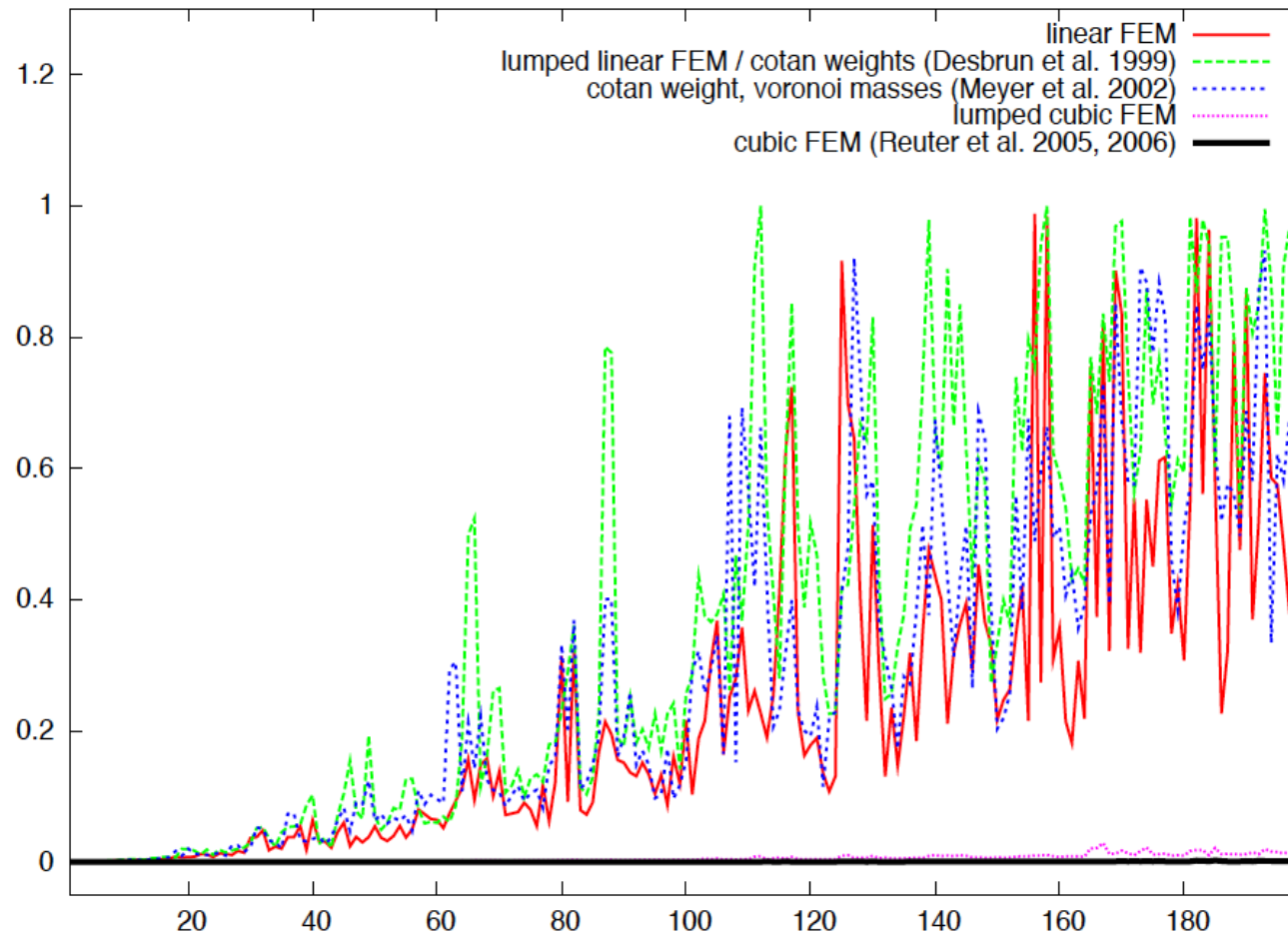


cubic

[SMI09]

+ Comparison Eigenfunctions

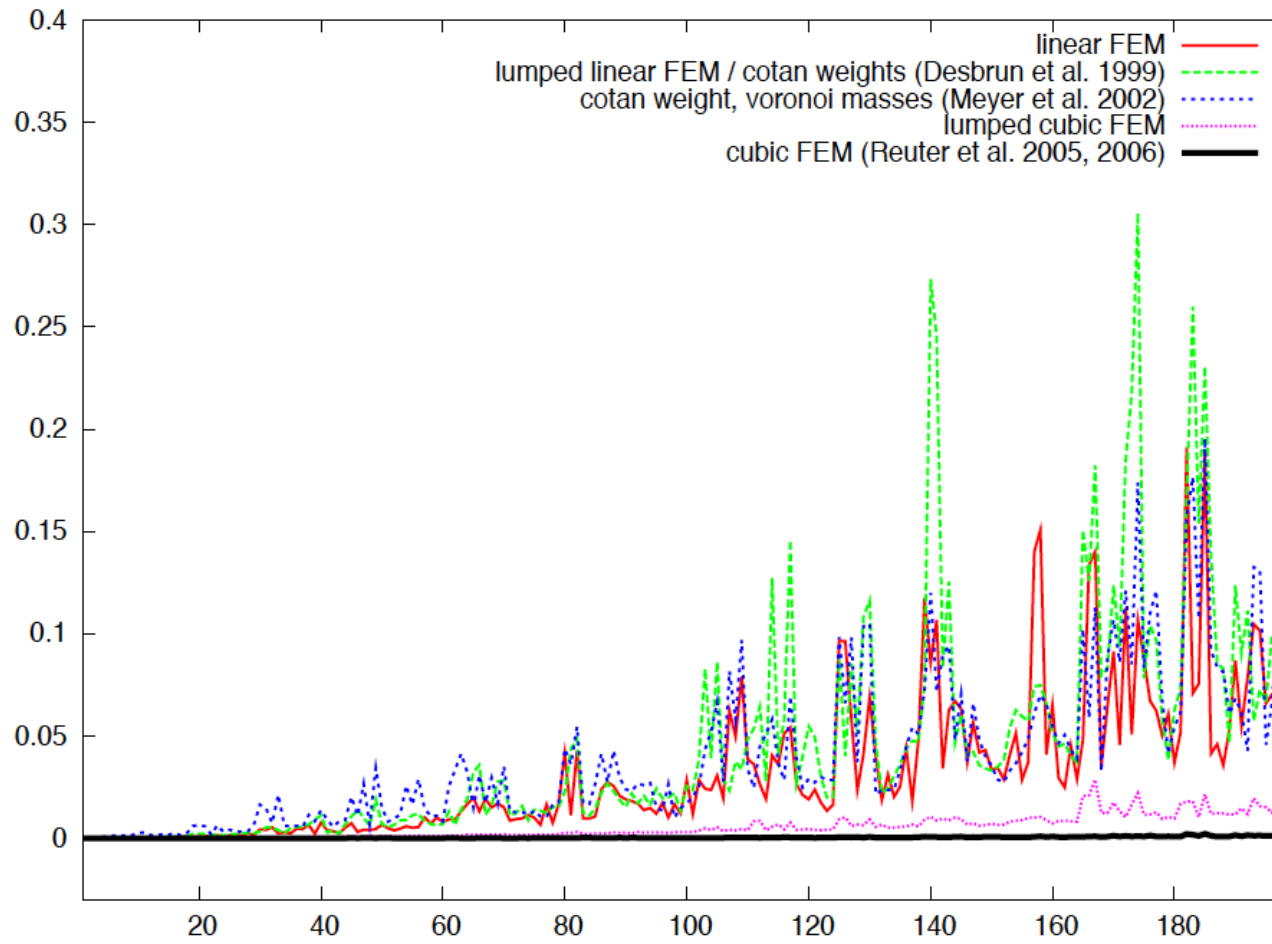
Rectangle - Uniform Mesh - first 200 Eigenfunctions:



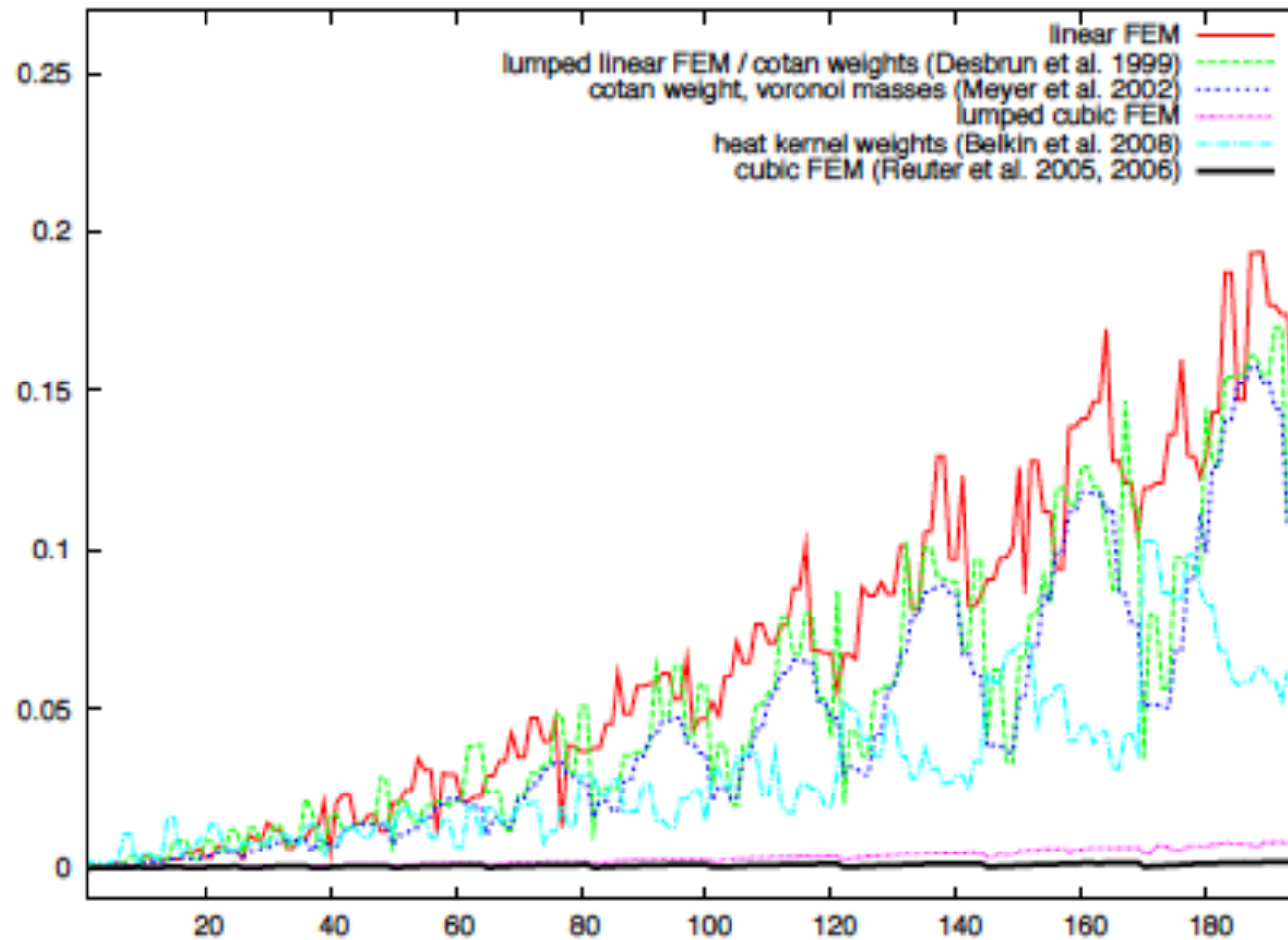
[SMI09]

+ Comparison Eigenfunctions

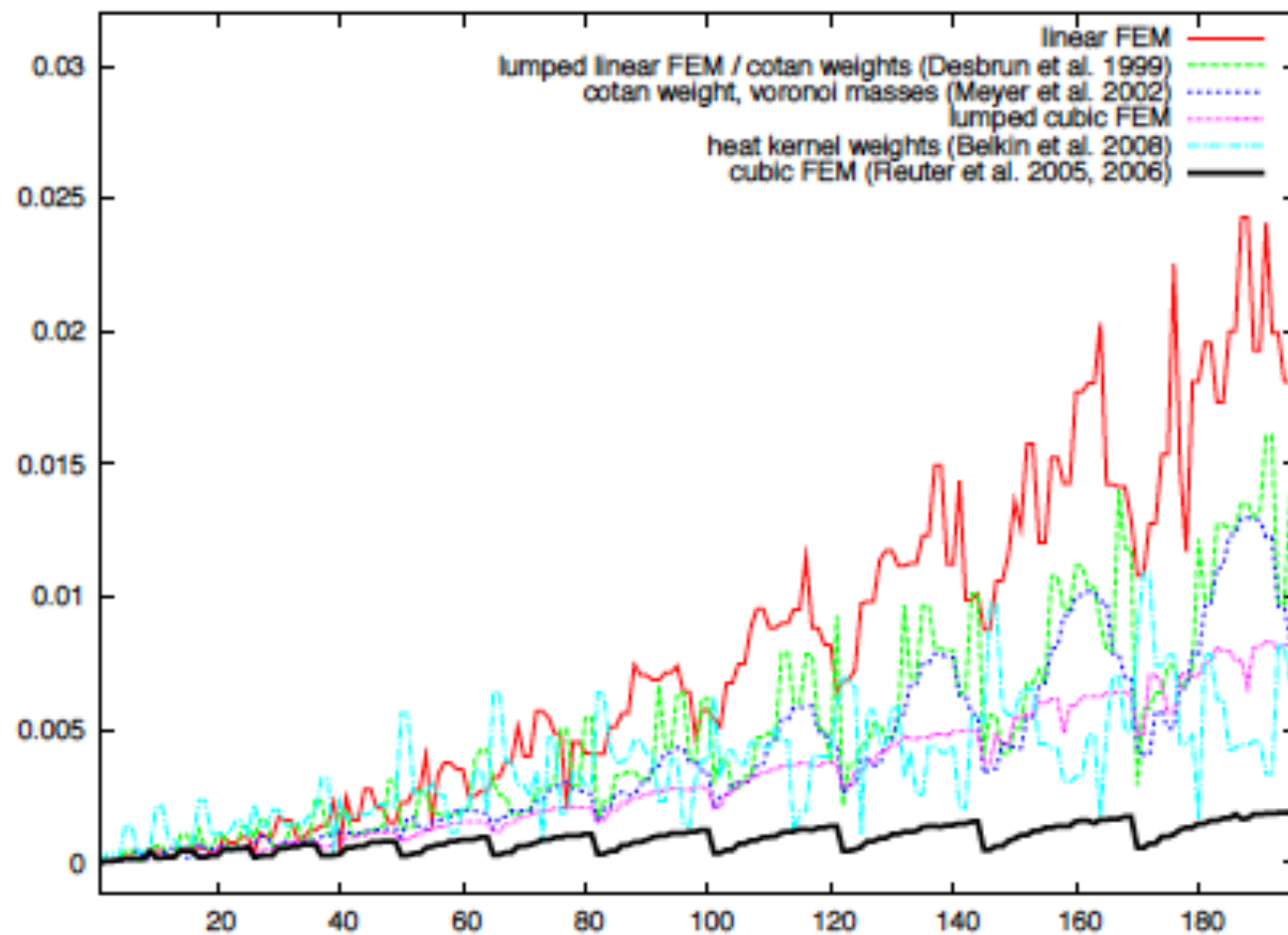
Rectangle - Uniform Mesh (same DOF as cubic) - 200 EF:



+ Comparison on the sphere



+ Sphere – same DOF



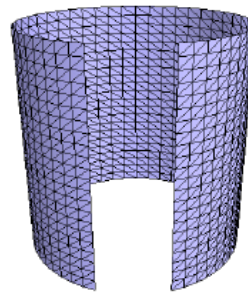
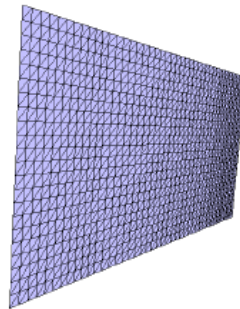
+ What is Shape and what is similar?

■ Shape should be invariant with respect to:

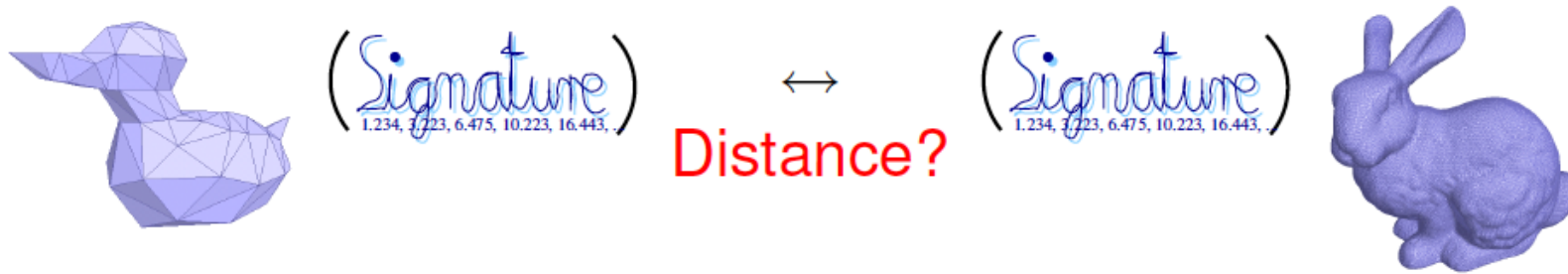
■ Location (rotation, translation)

■ Size

■ Isometries?



+ Shape Matching



- Prior alignment, scaling of the objects:
 - normalization, registration
- Computation of a simplified representation
 - Signature, Shape-Descriptor
- Comparison of the signatures
 - distance computation to measure similarity
- Disadvantages of current methods:
 - Over-simplification, missing invariance, complex pre-processing, difficult to compare signatures, support only special representations

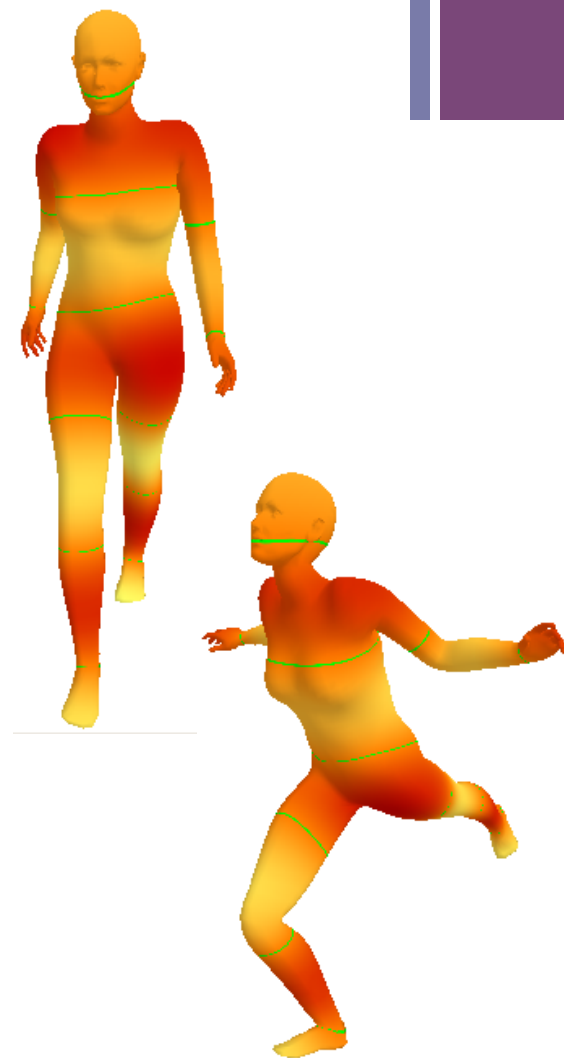
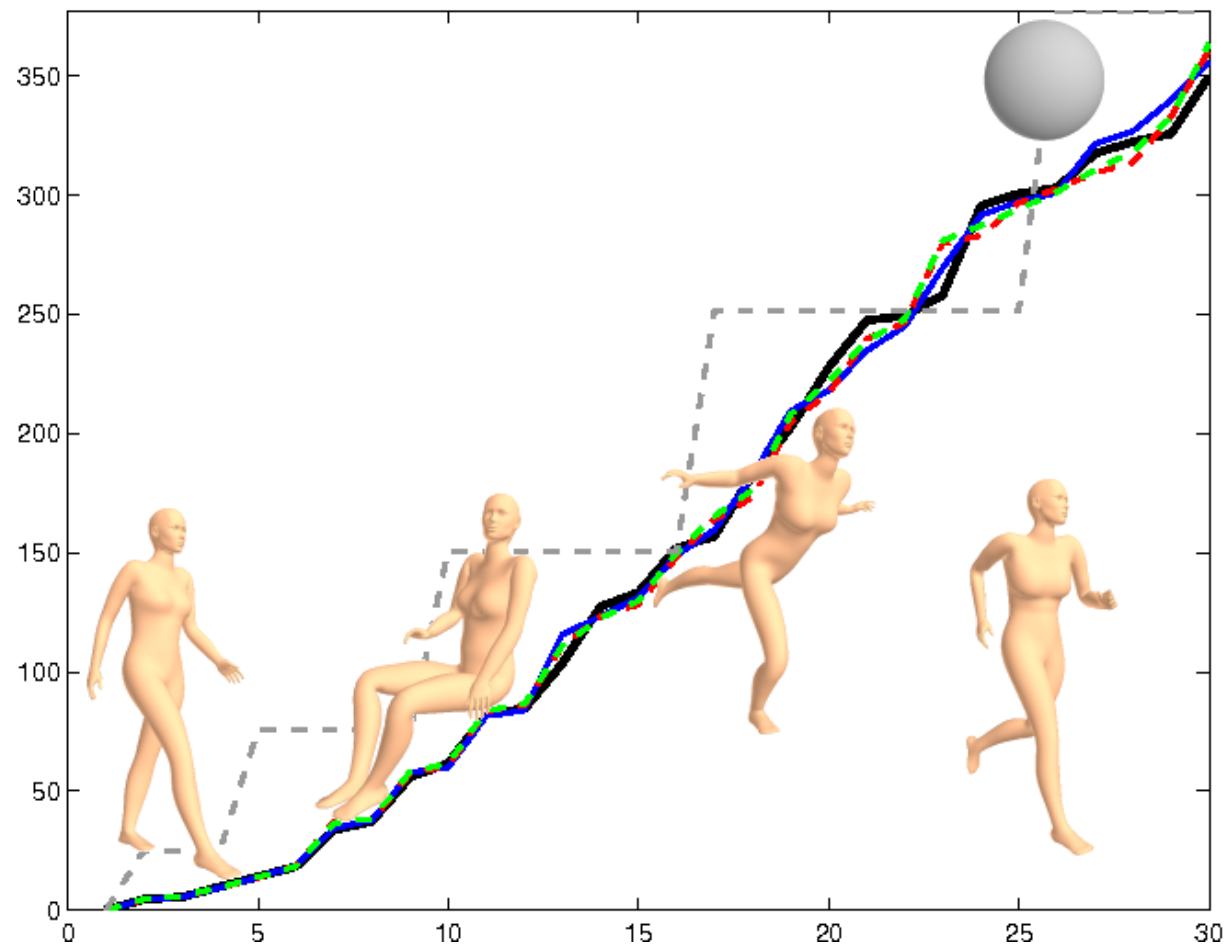
+ New Signature: ShapeDNA [spm05]

We use the (normed) n -dim vector of the **smallest n eigenvalues** $(\lambda_1, \dots, \lambda_n)$ of the Laplace operator Δ as the signature:



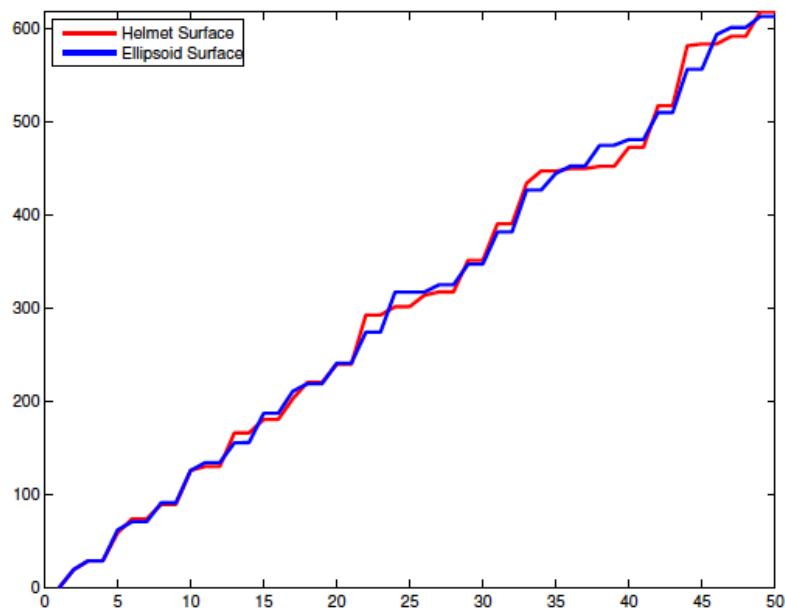
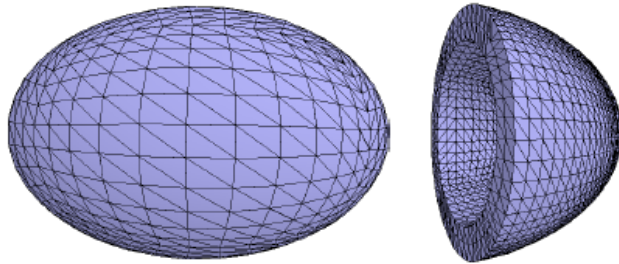
- Isometry invariant \Rightarrow location invariant
- and (where required) scaling invariant
- No registration necessary
- Surfaces & solids (even with cavities)
- Independent of representation
- Simple distance computation of the signature vectors
- Efficient and highly accurate computation with cubic FEM

+ Isometry Invariance

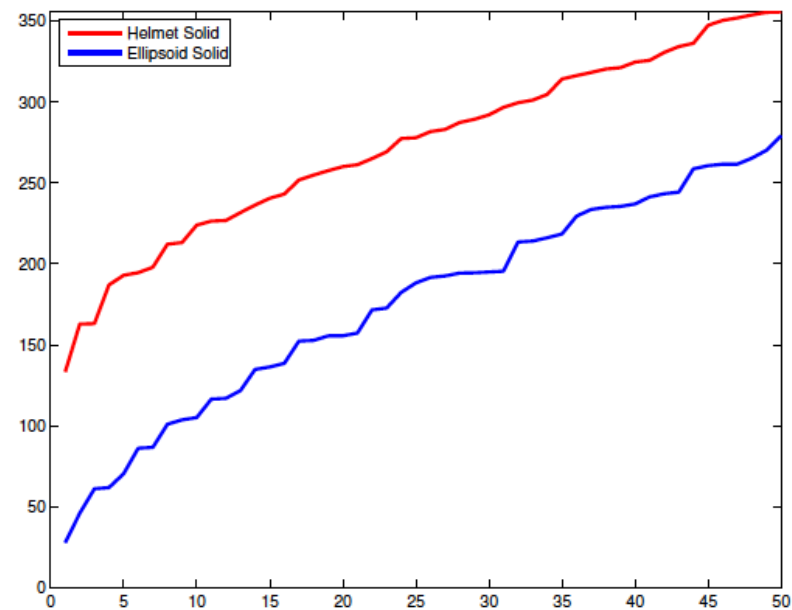
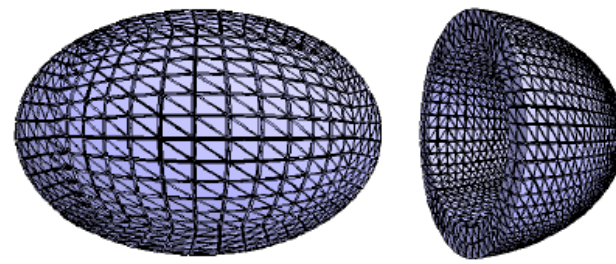


+ 2D near isometry, 3D not

2D Surface ShapeDNA

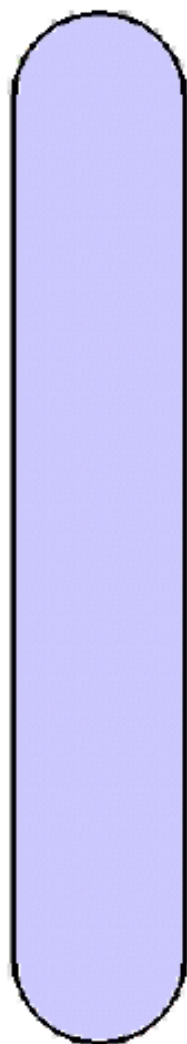


3D Solid ShapeDNA

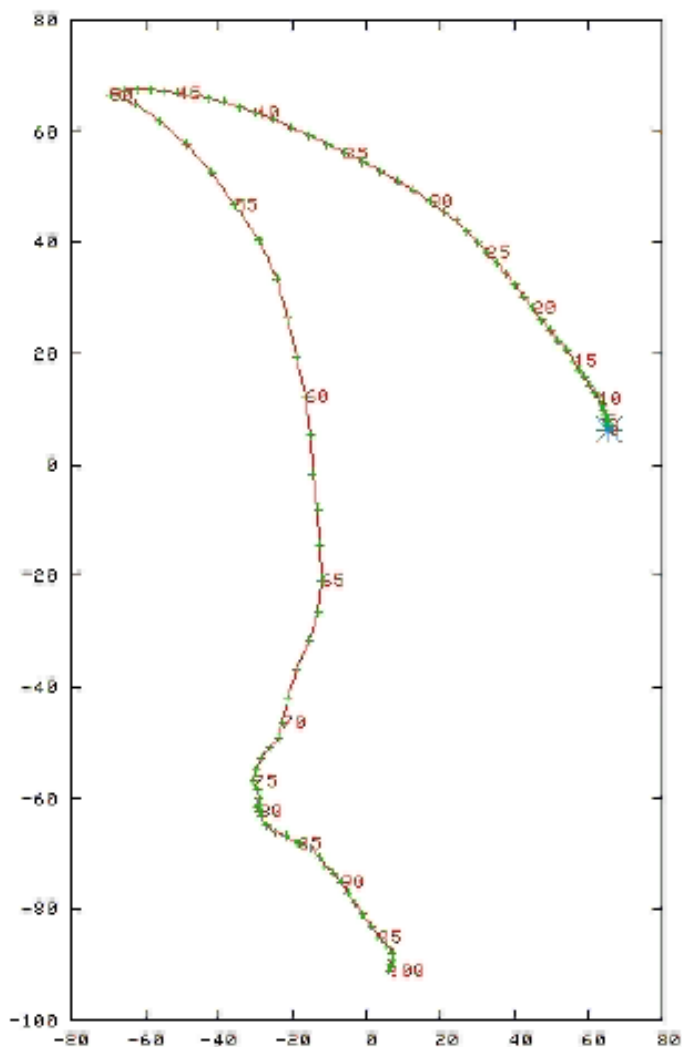


For solid bodies in \mathbb{R}^3 isometry is equivalent to congruency.

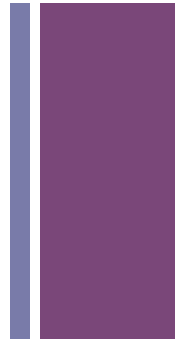
+ Continuous Shape Dependence



M.Reuter

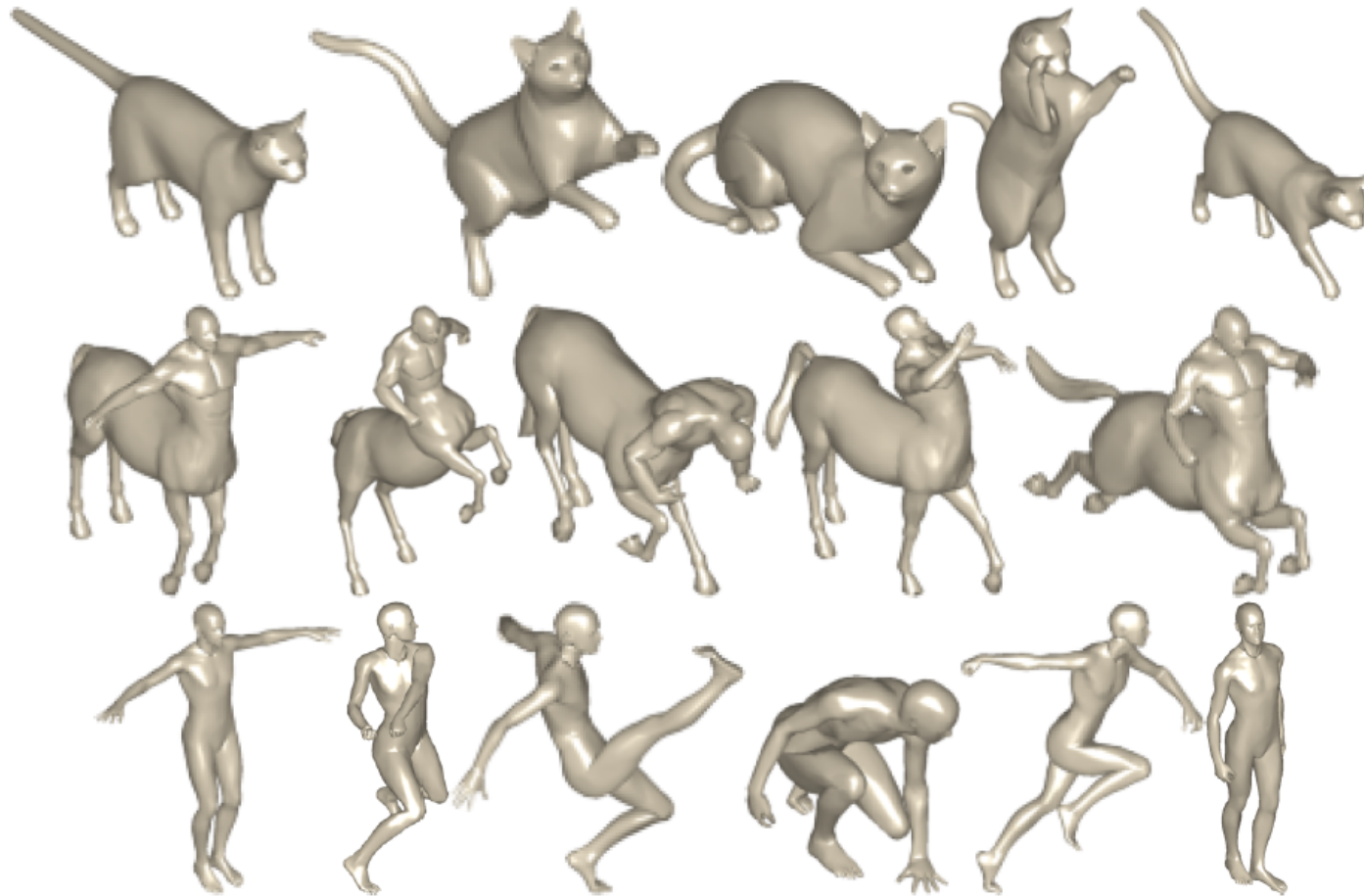


+ Database Retrieval



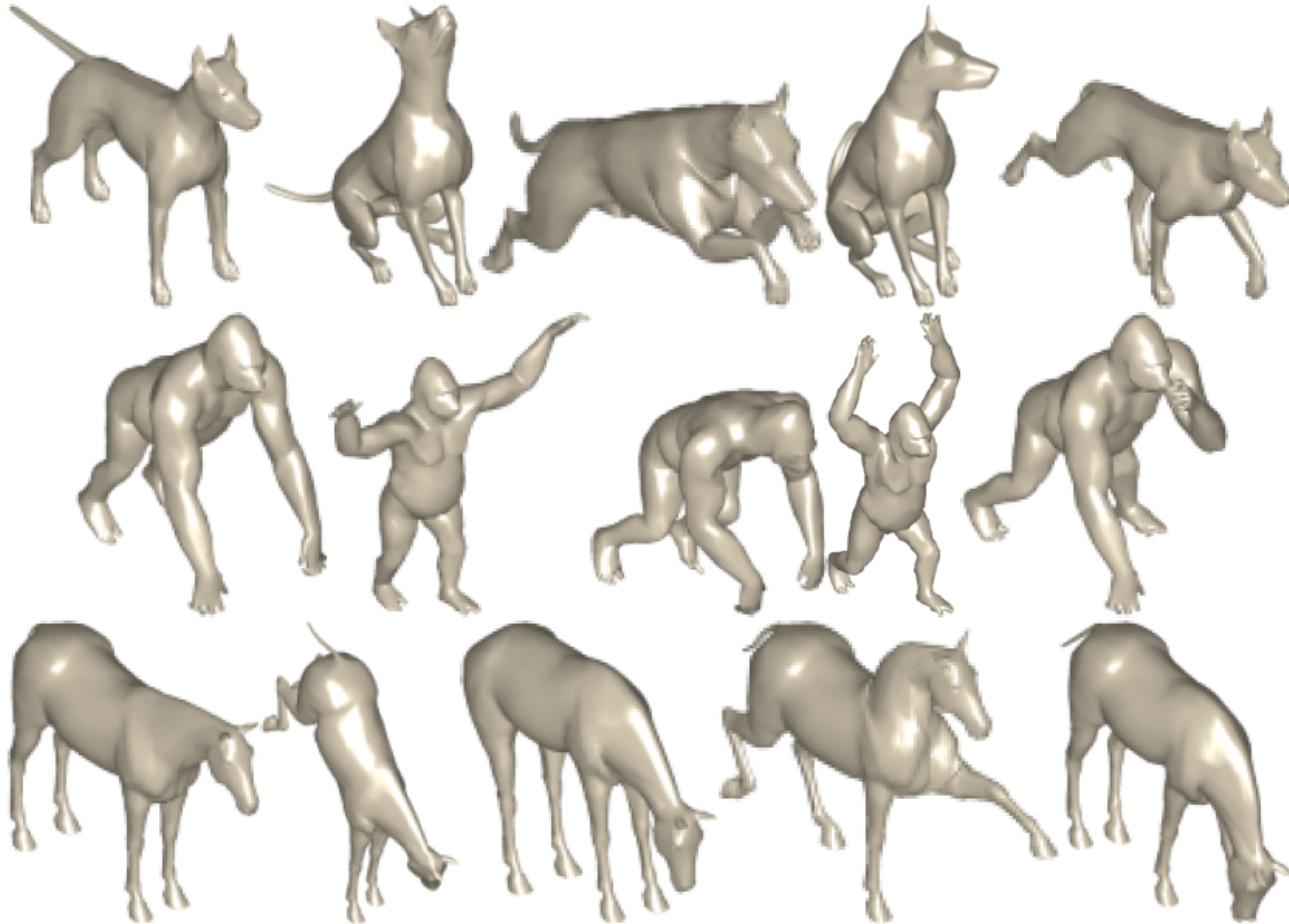
- 1. Computation of the first n Eigenvalues (Shape-DNA)
- 2. Normalization
 - a) Surface area normalized
 - b) Volume normalized
- 3. Distance computation of the Shape-DNA (n -dim vector)
 - a) Euclidean distance (\leftarrow default)
 - b) Another p -norm
 - c) Hausdorff distance
 - d) Correlation ...

+ Nonrigid Shape Database (148)



Courtesy of Bronstein, Bronstein, Kimmel, 2006

+ Nonrigid Shape Database (148)



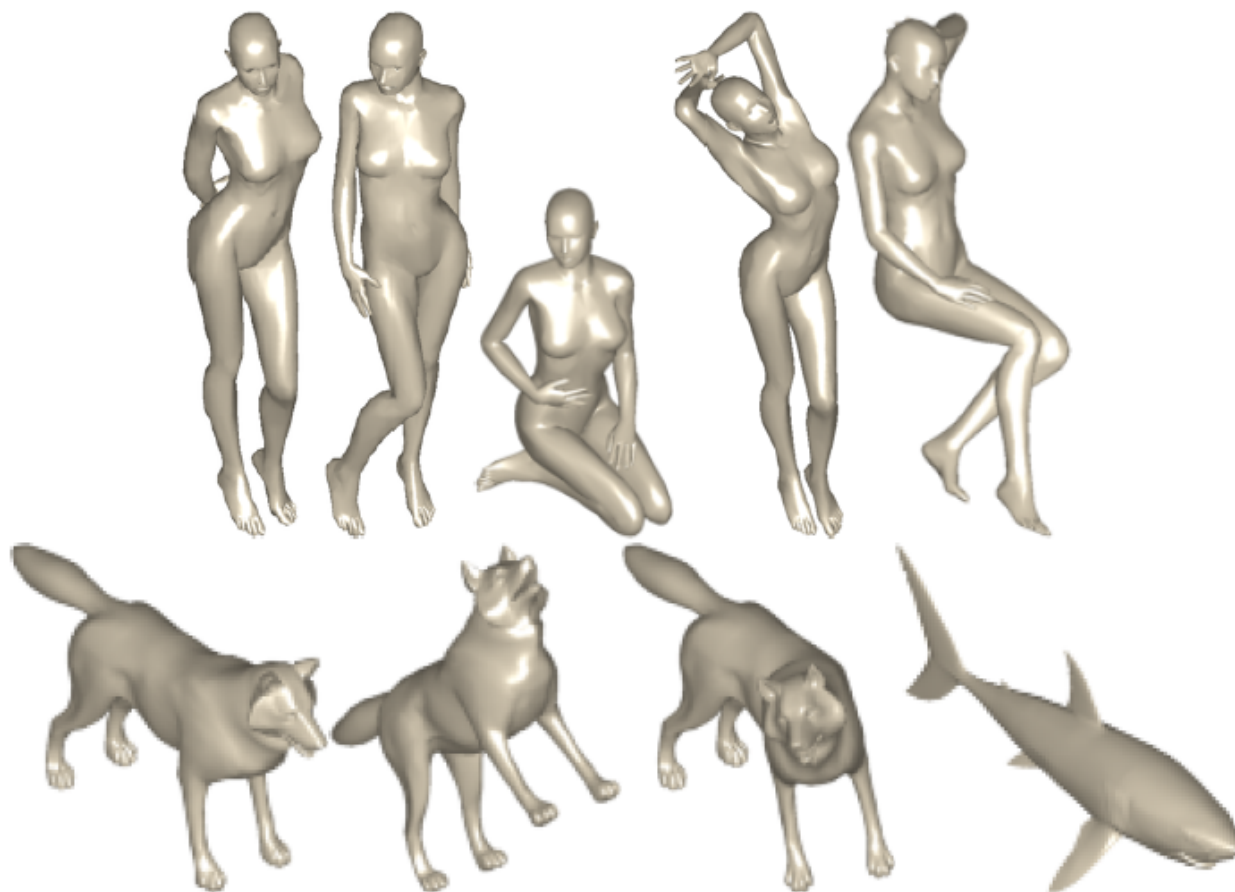
Courtesy of Bronstein, Bronstein, Kimmel, 2006

+ Nonrigid Shape Database (148)



Courtesy of Bronstein, Bronstein, Kimmel, 2006

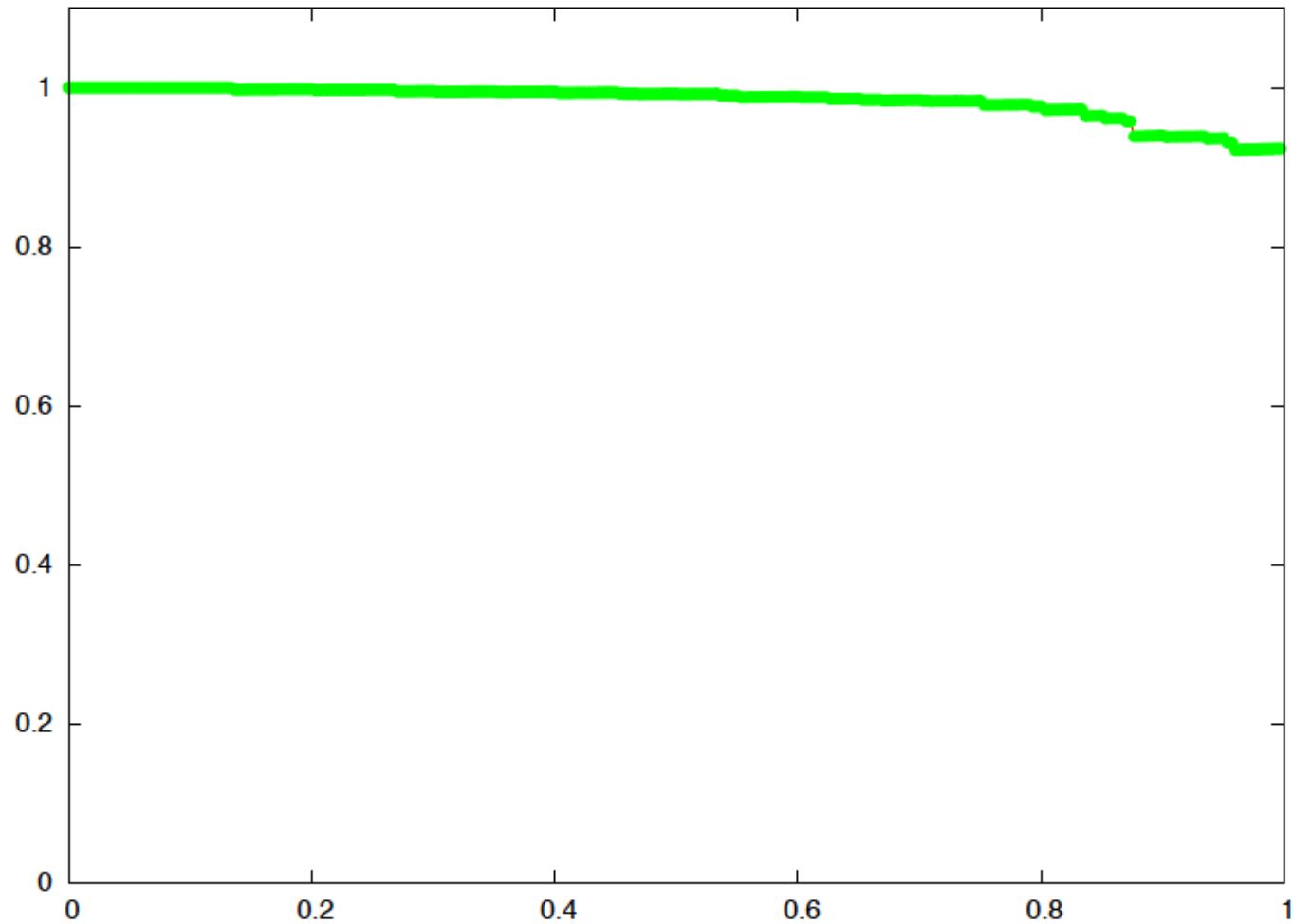
+ Nonrigid Shape Database (148)



Courtesy of Bronstein, Bronstein, Kimmel, 2006

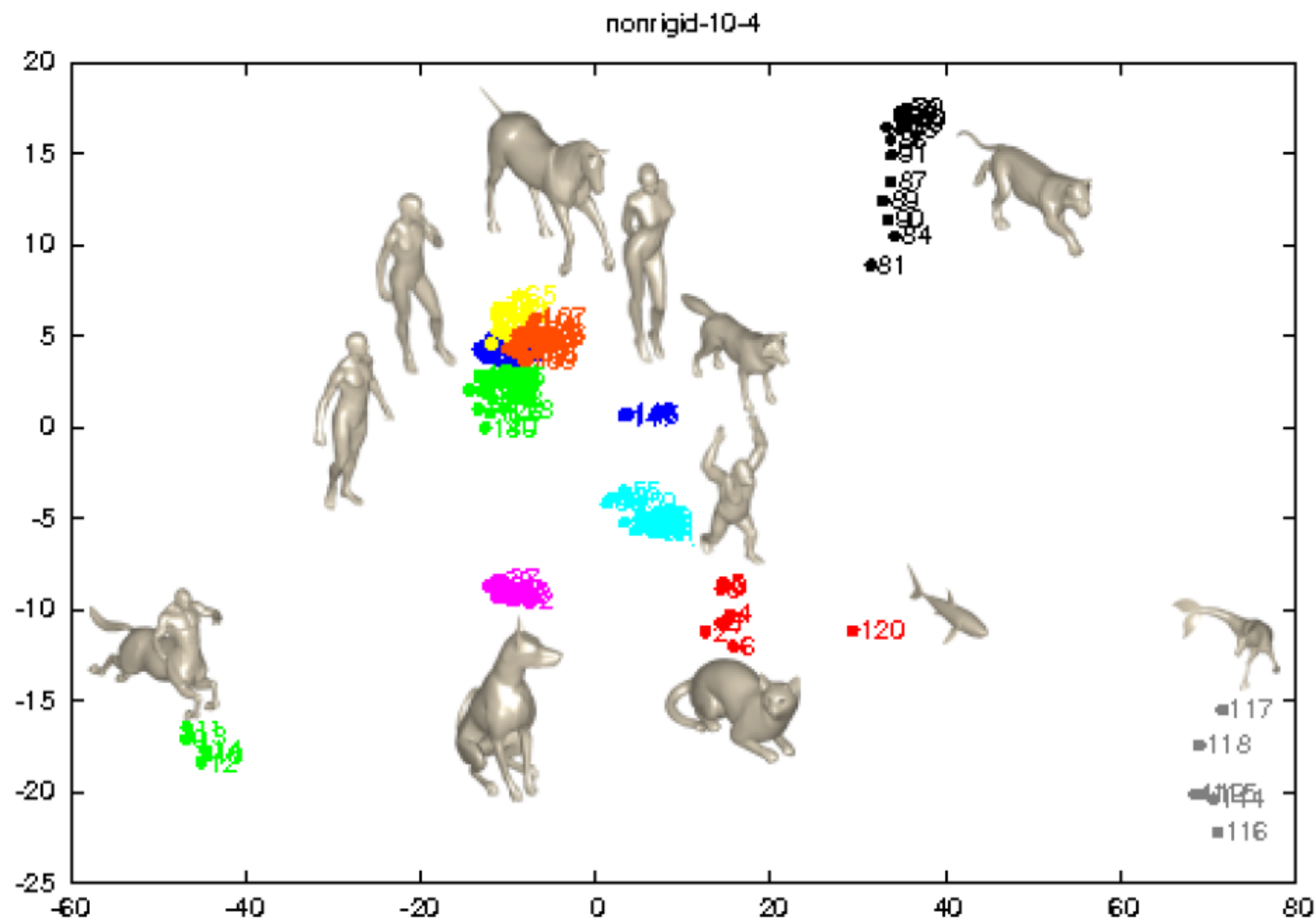
+ Nonrigid Shape Database (148)

$P(R) := \text{Precision}(\text{Recall})$ averaged:

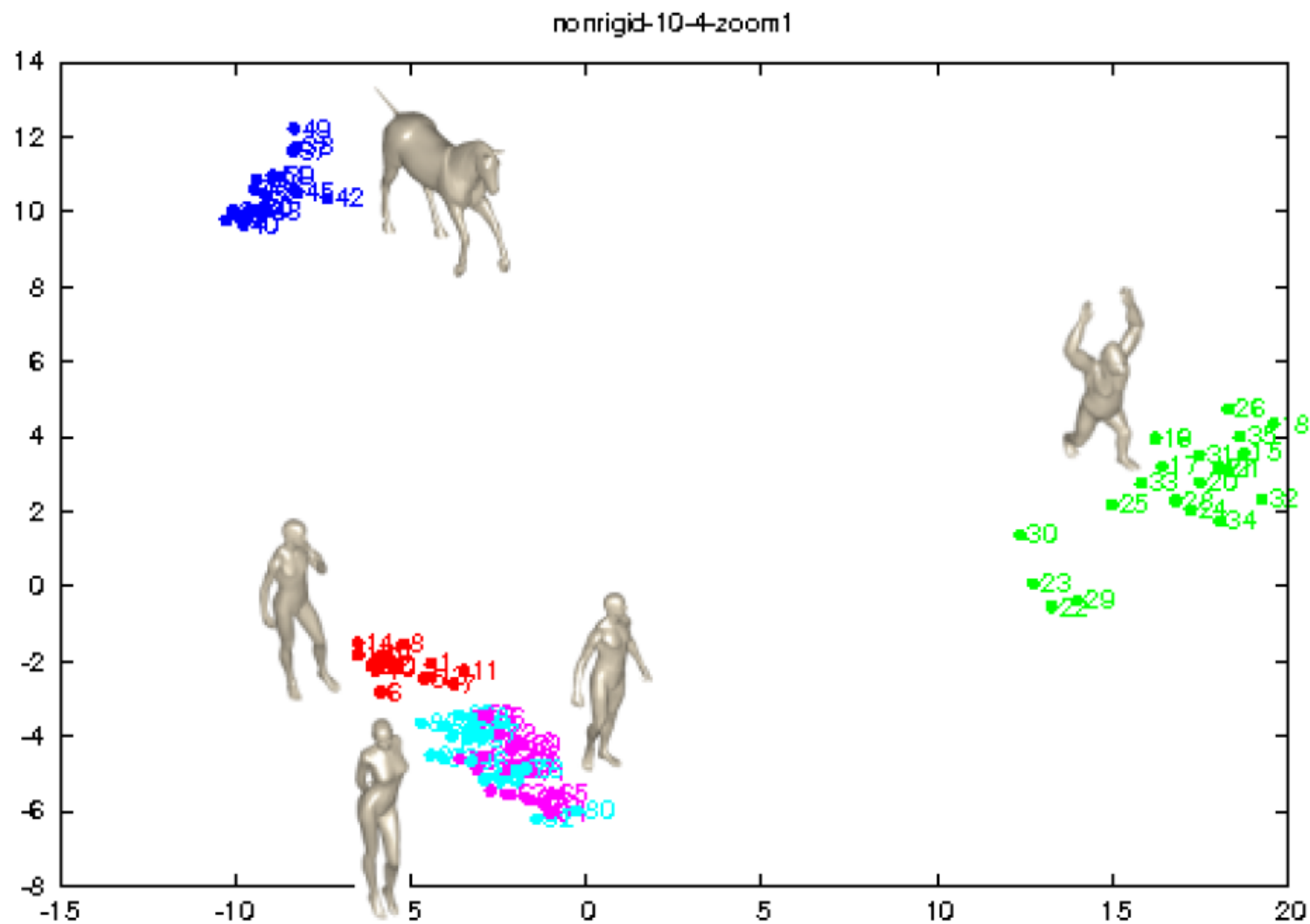


Integral: 0.983301 and min: 0.921847

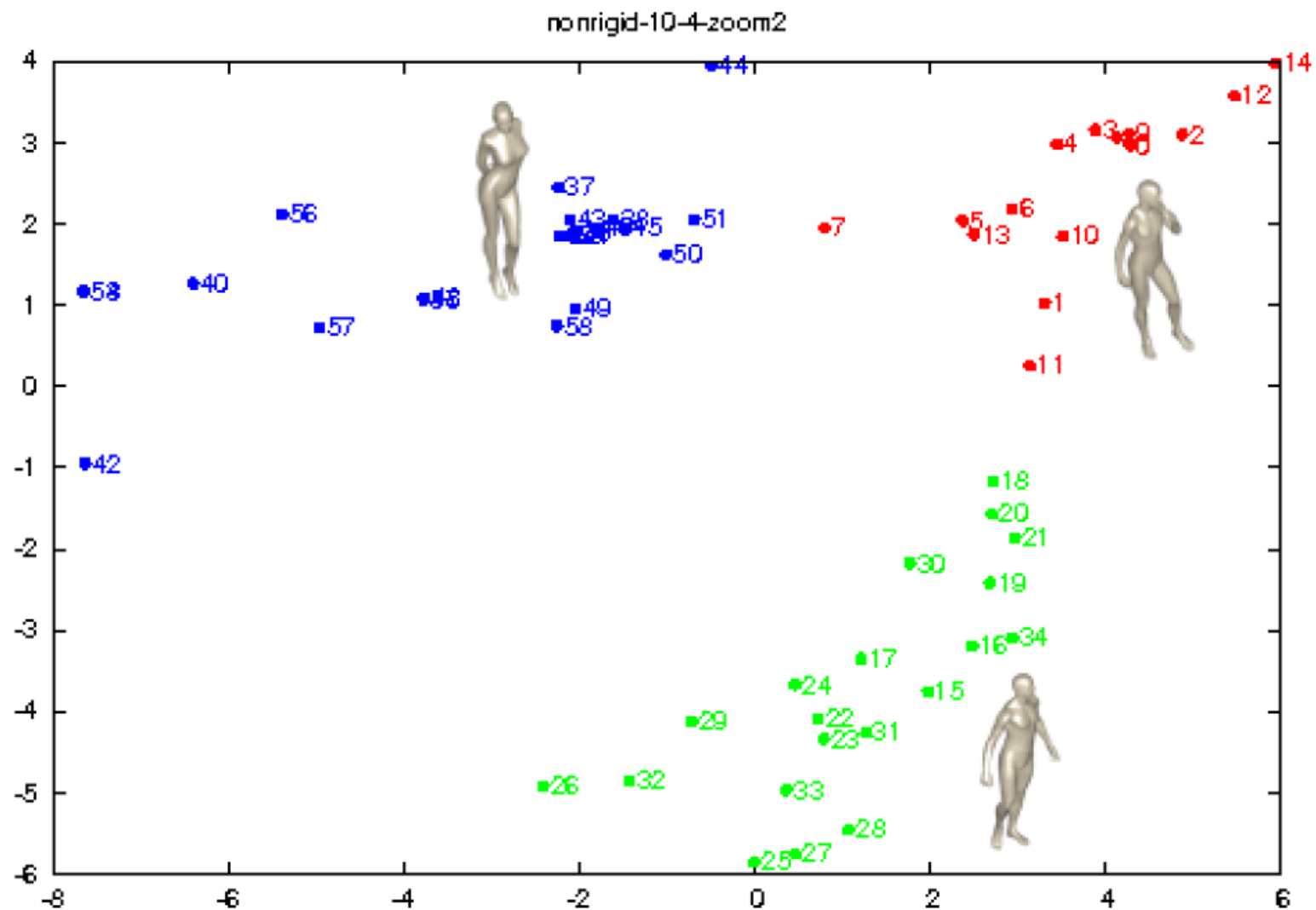
+ Nonrigid DB – MDS Plot



+ Nonrigid DB – Zoom 1

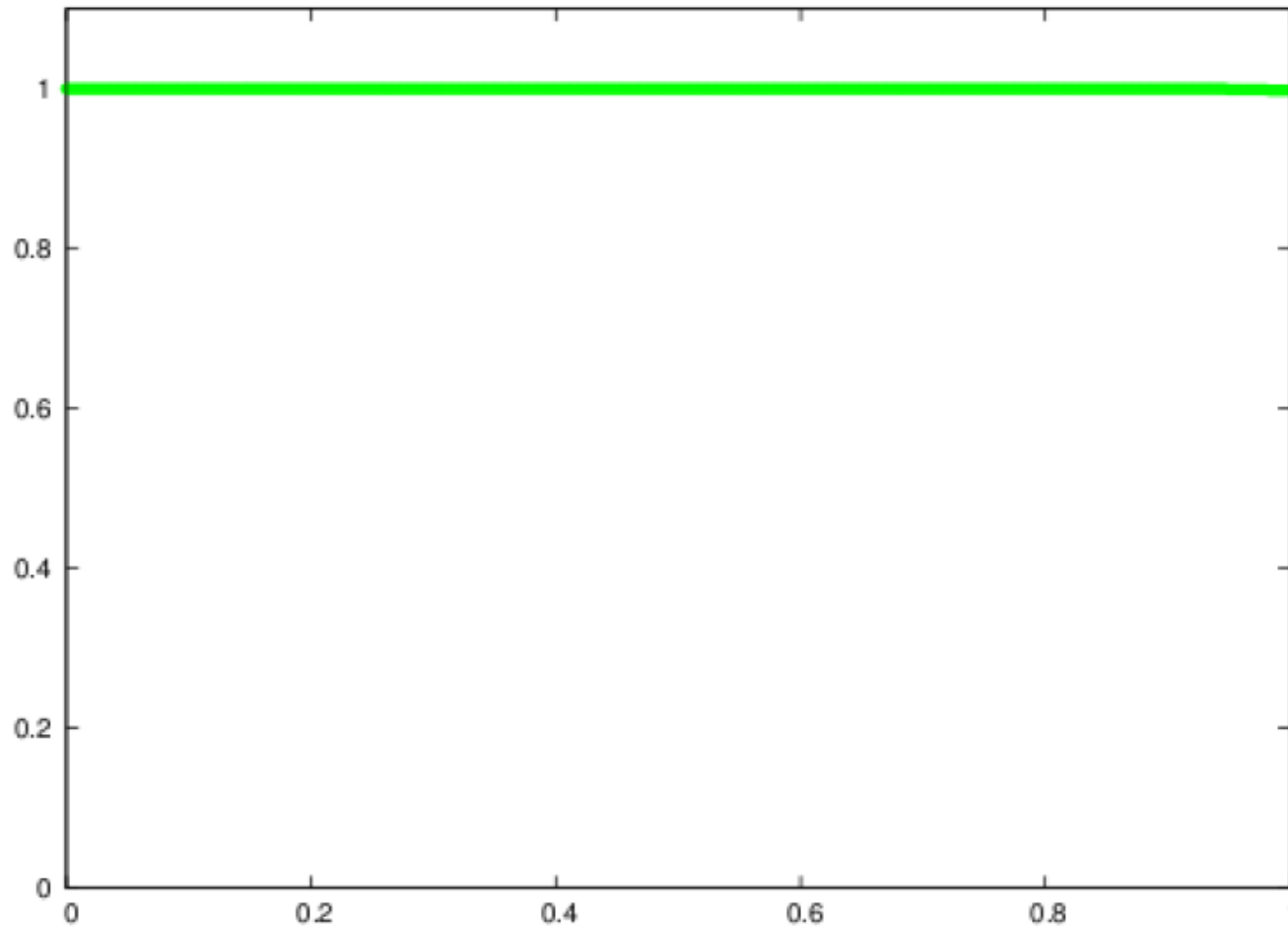


+ Nonrigid DB – Zoom 1



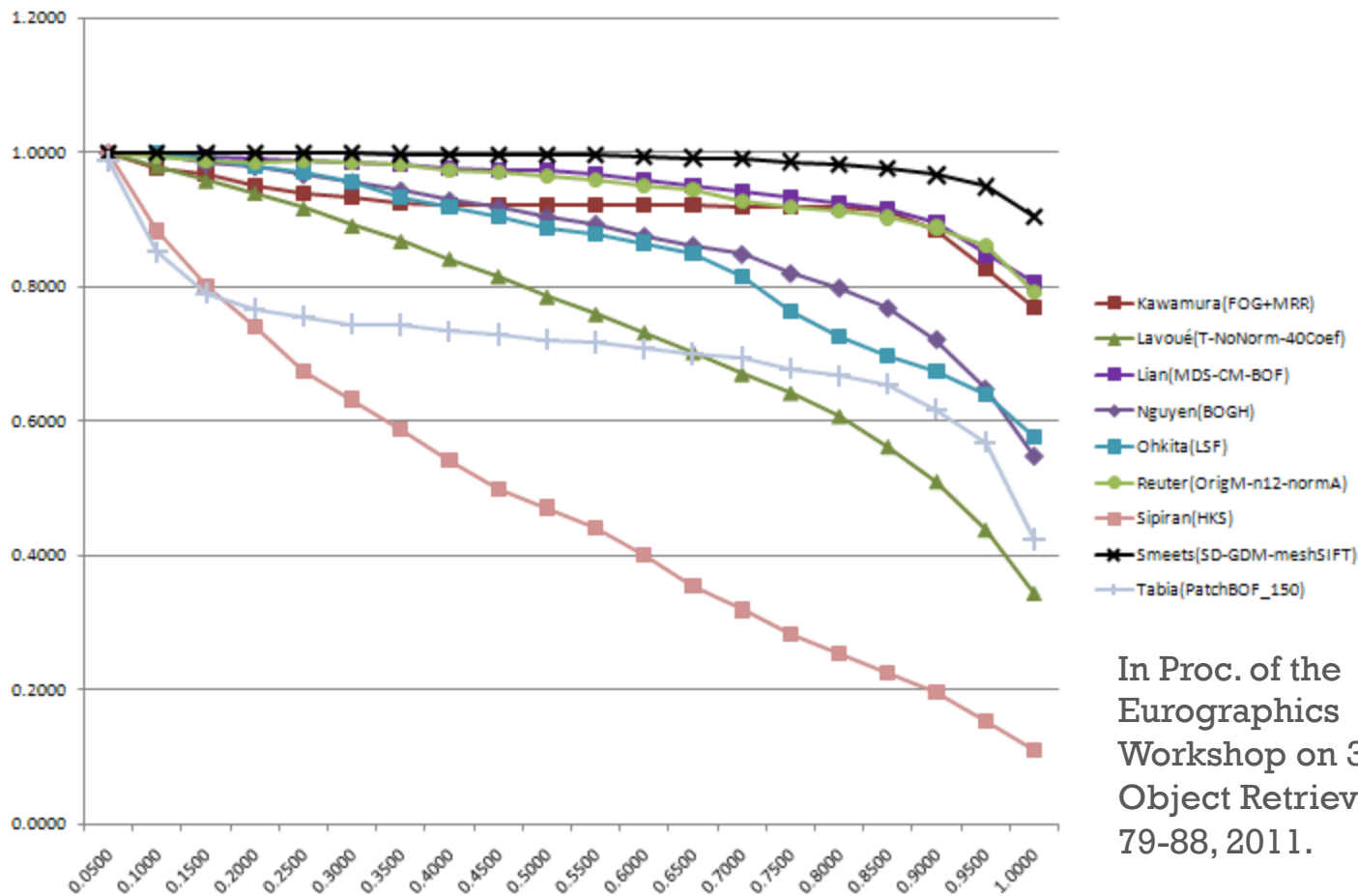
+ Nonrigid DB

$P(R) := \text{Precision}(\text{Recall})$ averaged:



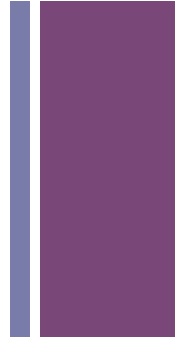
Integral: 0.999947 and min: 0.99829

+ Shape Retrieval Contest 11 Non Rigid Track



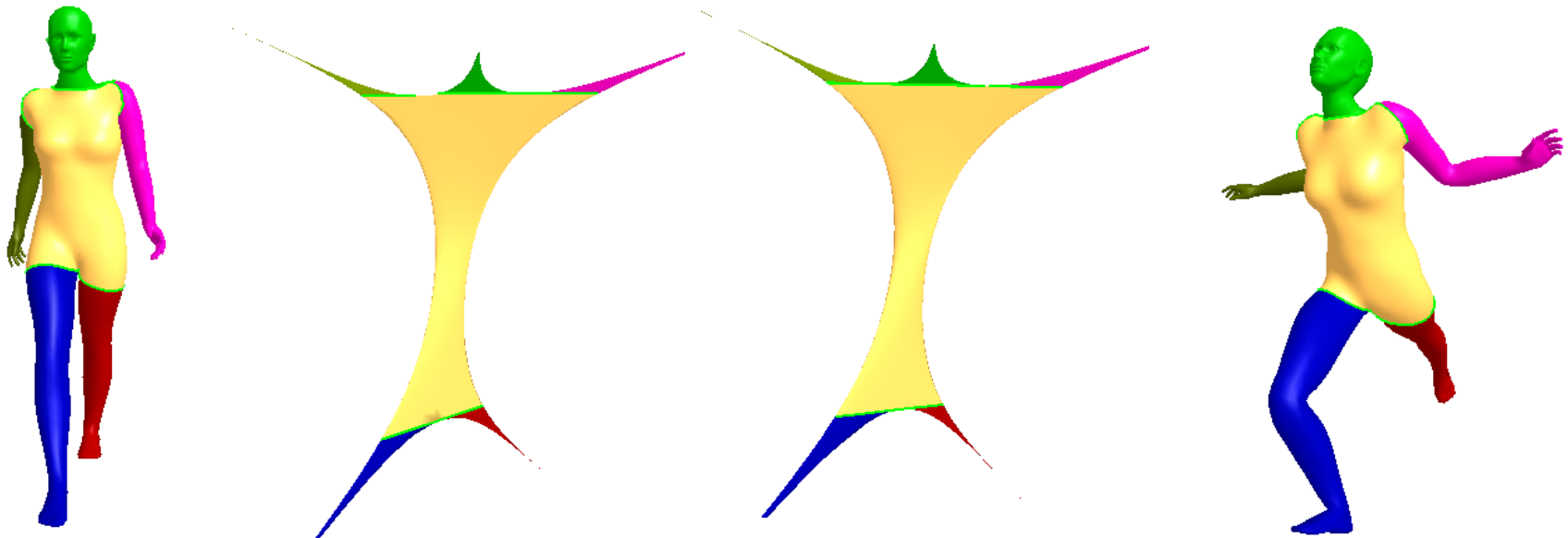
In Proc. of the
Eurographics
Workshop on 3D
Object Retrieval, pp.
79-88, 2011.

+ Spectral Embedding

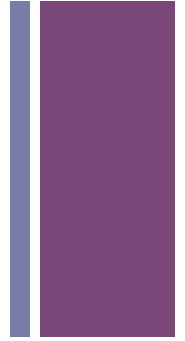


- Maps each vertex to the value of all (or a few) Eigenfunctions at that location:

$$\Phi(\vec{x}) = (f_1(\vec{x}), f_2(\vec{x}), f_3(\vec{x}), \dots)$$

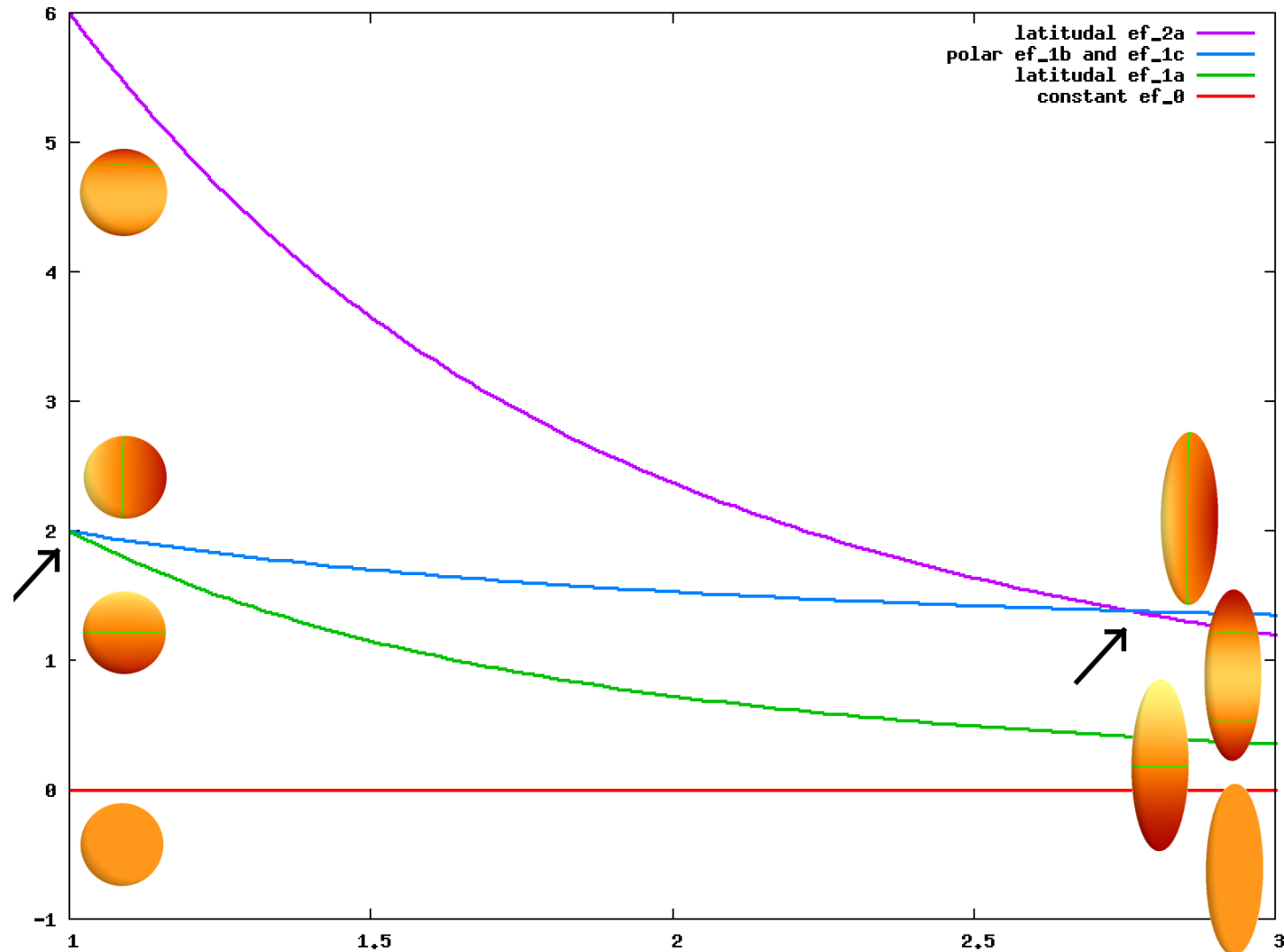


+ Difficulties

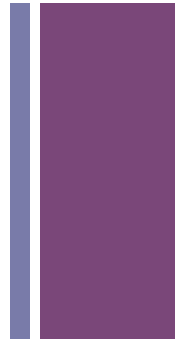


- Numerical inaccuracies
- Sign Flips
 - Scalar multiples are contained in same space
- Higher dimensional Eigenspaces
 - Arbitrary basis (luckily they are rare, but being close to one is already problematic)
 - Due to numerical errors Eigenvalues will be slightly different and we cannot really detect these situations
- Switching of Eigenfunctions
 - Occurs, because of numerical instabilities (two close Eigenvalues switch)
 - Or due to geometric (non-isometric) deformations

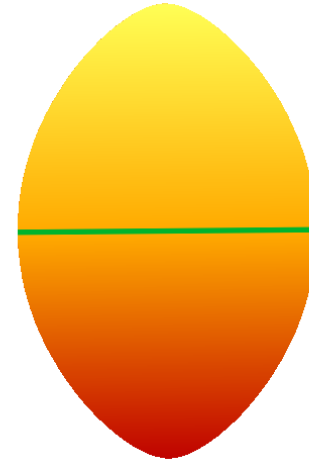
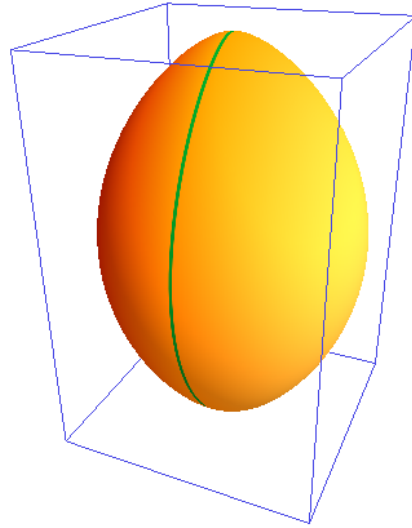
+ Switching of Eigenfunctions



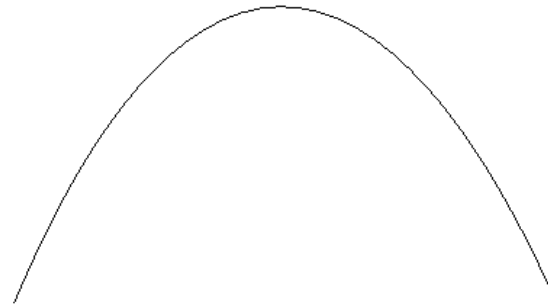
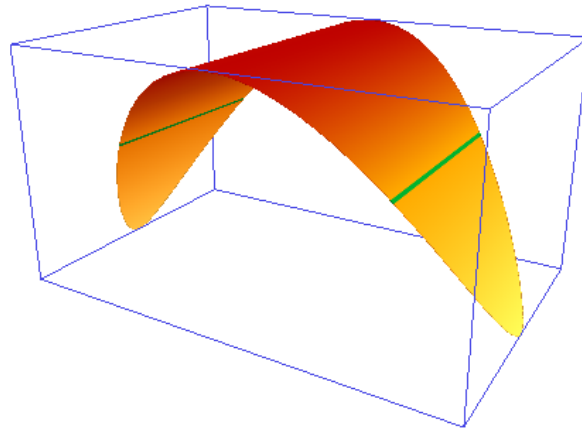
+ Switching of Eigenfunctions



■ $Z=2.74$

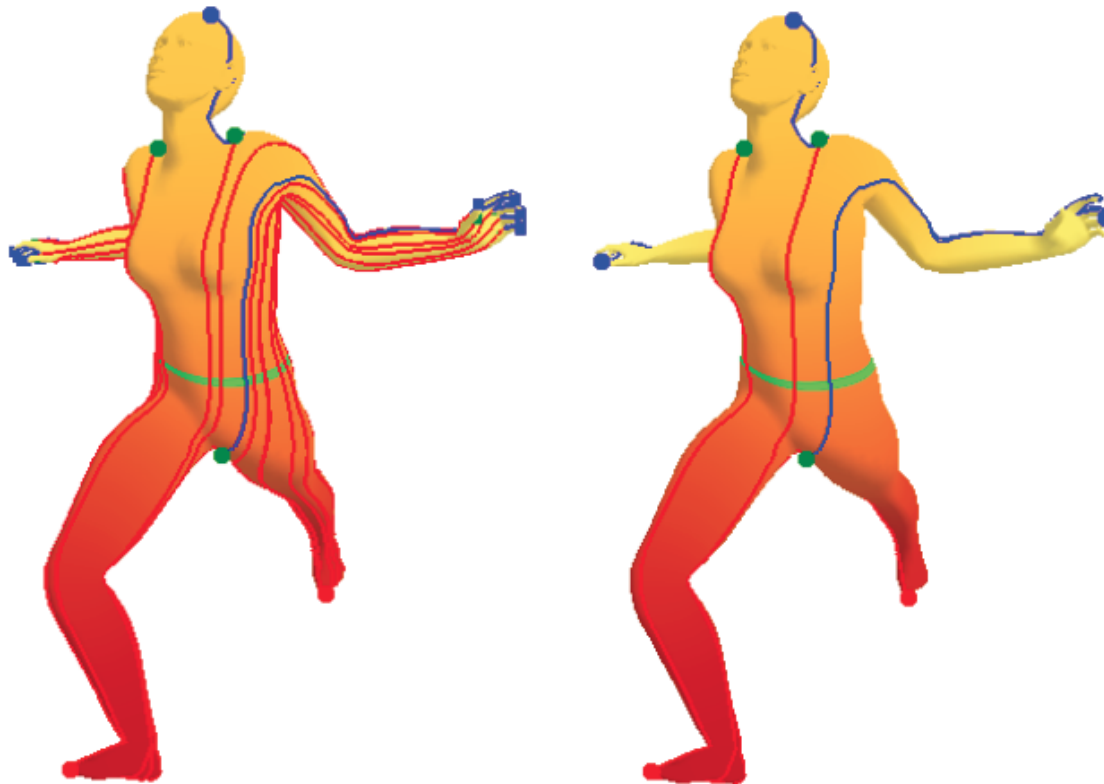


■ $Z=2.76$



+ Shape Segmentation

- Morse-Smale Complex of the 1st Eigenfunction
 - Left: full complex Right: simplified (3min,2max,3saddles)



[IJCV09]

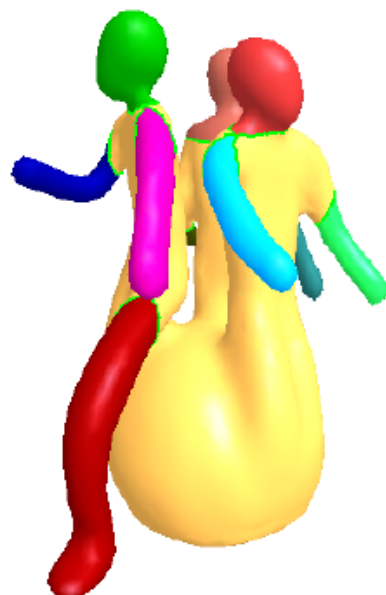
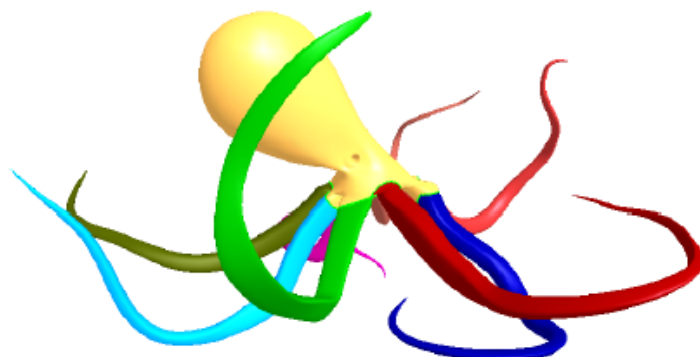
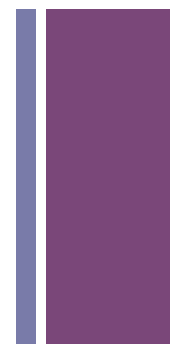
+ Shape Segmentation

- Segmentation on different 'persistence' levels
 - Left: using only the most significant critical points
 - Right: close-up of hand using all (except noise)



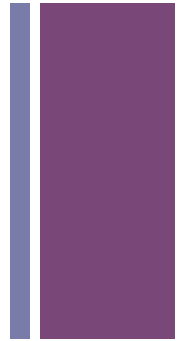
[IJCV09]

+ Hierarchical Segmentation



[IJCV09]

+ Consistent Segmentation and Registration



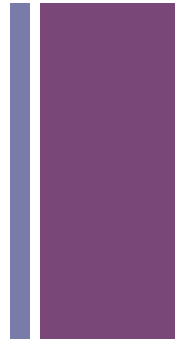
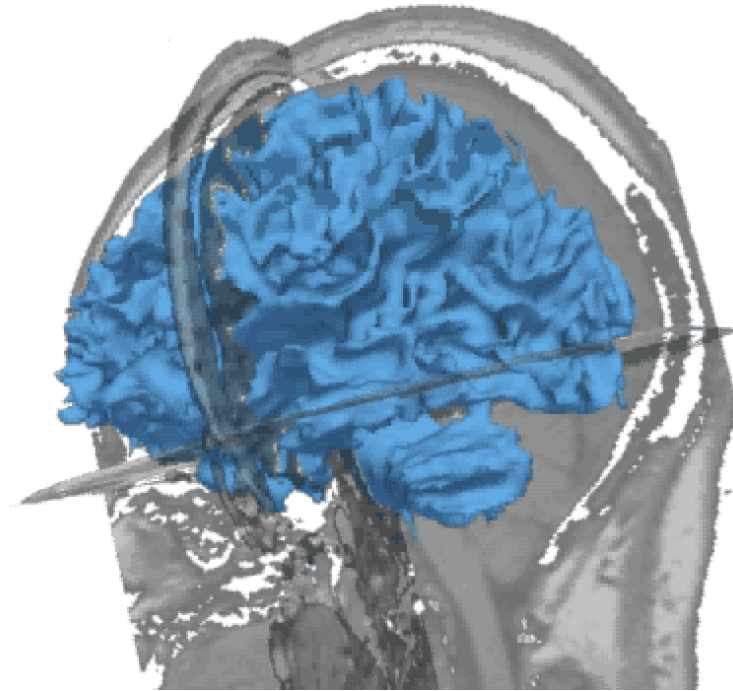
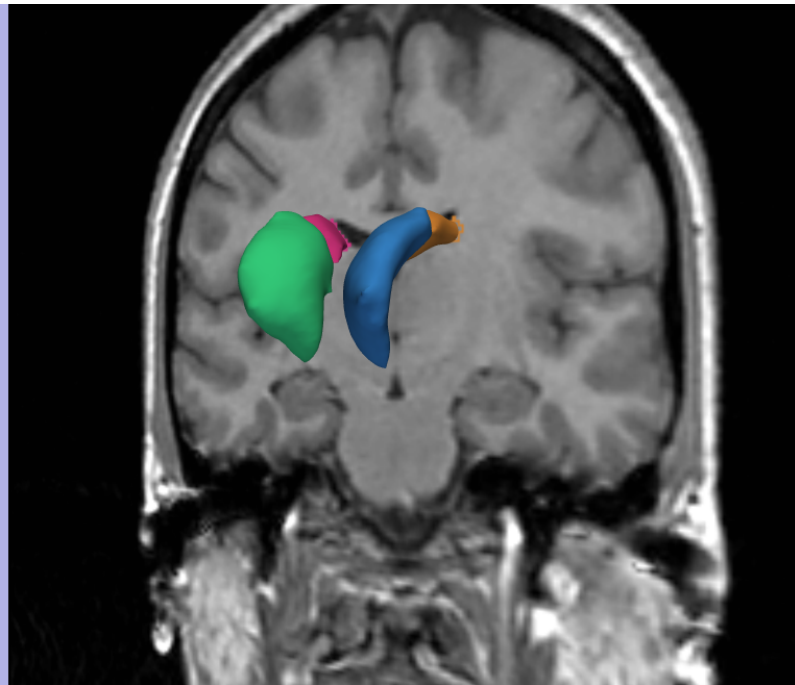
[IJCV09]

+ Other Applications

- Dense correspondence: Texture or Marker transfer, Surgical Planning
- Segmentation plus Skeleton: Pose Interpolation, Animation



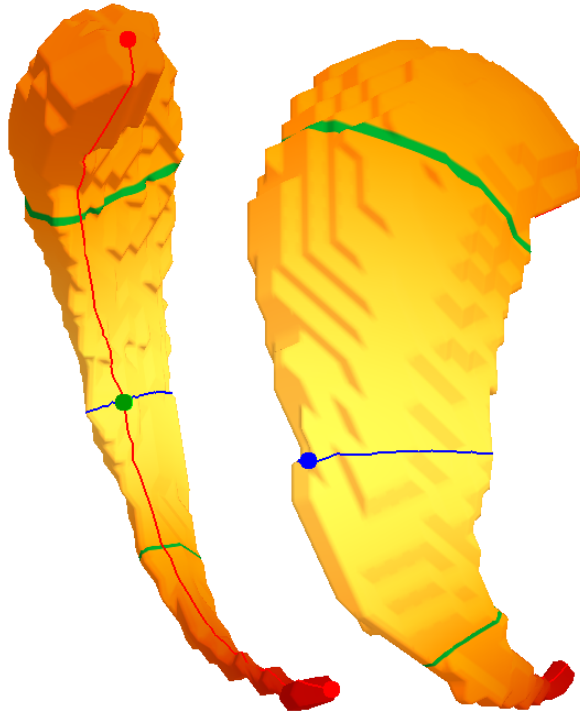
+ Caudate Nucleus



- Involved in memory function, emotion processing, and learning
 - Psychiatry Neuroimaging Lab (BWH - Martha Shenton)
 - Population: 32 Schizotypal Personality Disorder, 29 NC

[MICCAI07], [CW08], [CAD09]

+ Shape Analysis Caudate

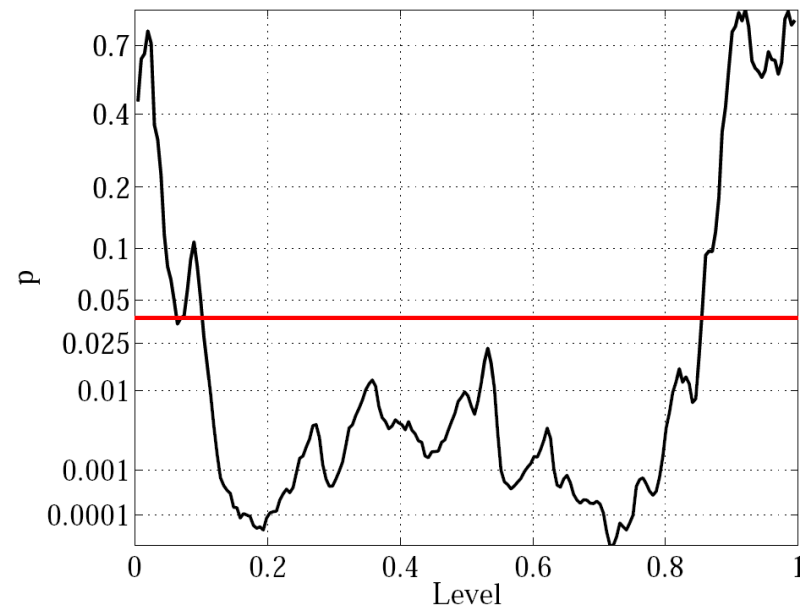
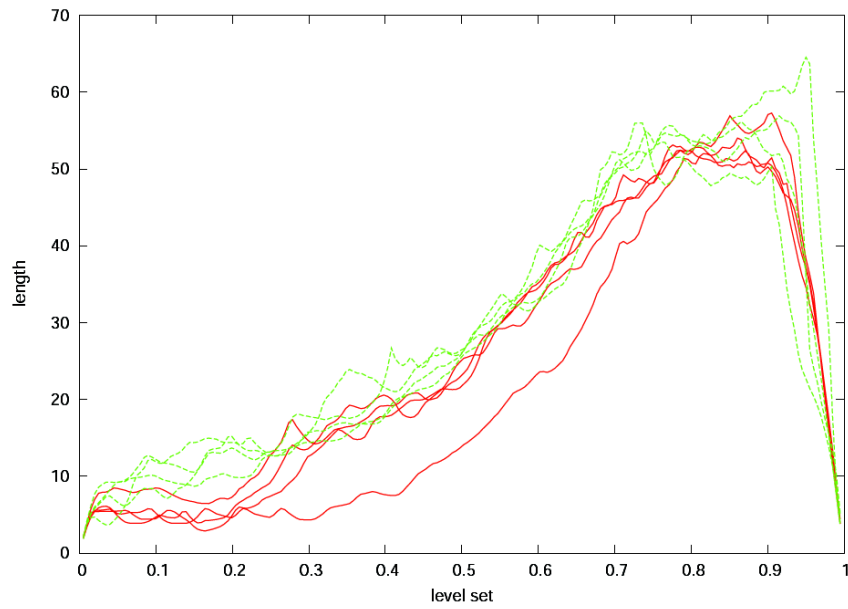
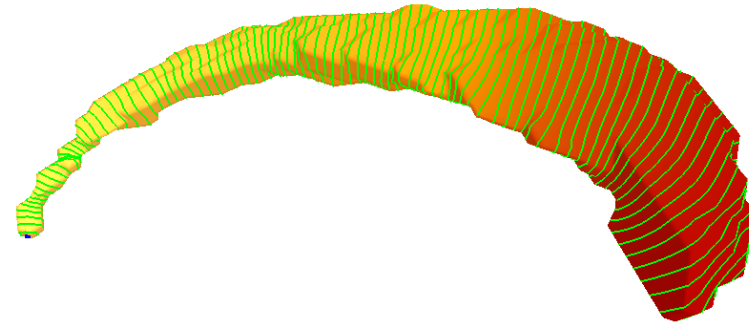
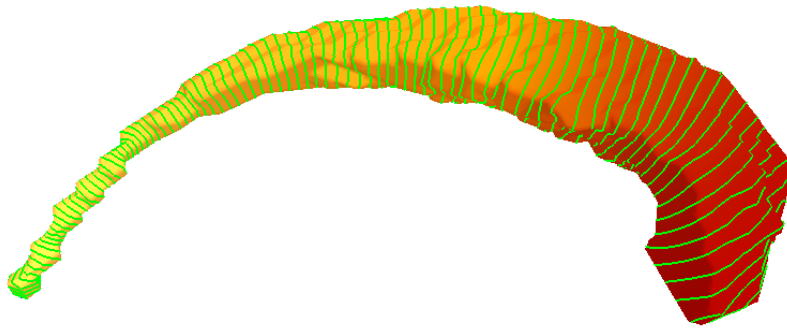
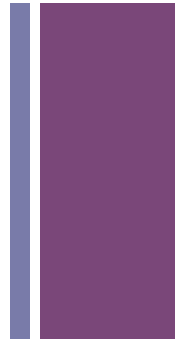


- Eigenfunction (EF): 2
- maxima at tips (red)
- minimum at outer rim (blue, middle)
- saddle at inner rim (green, left),
- integral lines (red and blue curve) run from the saddle to the extrema
- closed green curves denote the zero level sets

- (h) the head circumference (long green curve)
- (w) the waist circumference (blue curve)
- (t) the tail circumference (short green curve)
- (l) the length (red curve).

[MICCAI07], [CW08], [CAD09]

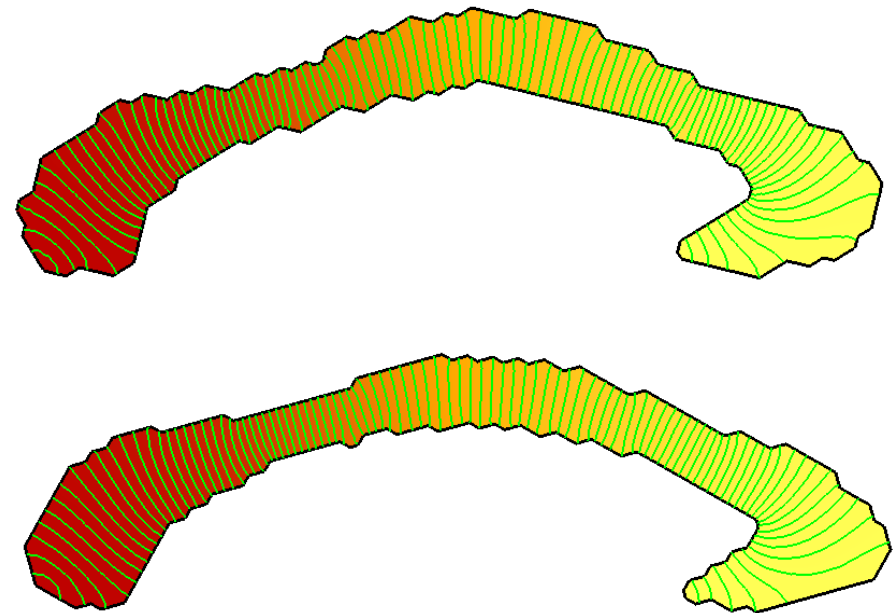
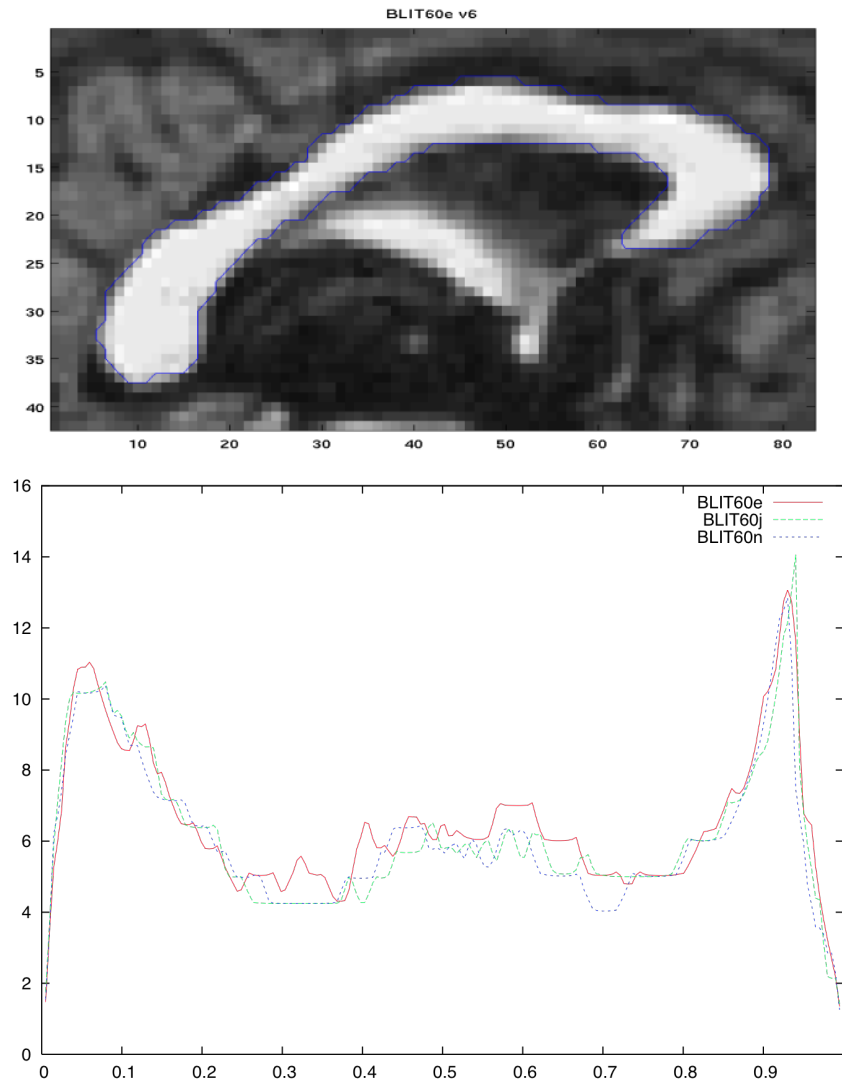
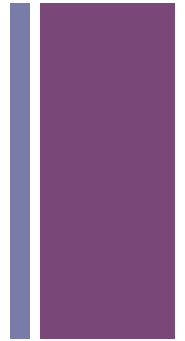
+ Shape Analysis Caudate



[MICCAI07], [CW08], [CAD09]



Shape Analysis Corpus Callosum



+ Thanks



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<http://reuter.mit.edu>

