

# Statistics of the Geometry of Object Populations

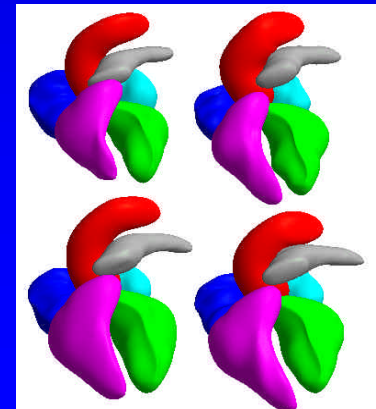
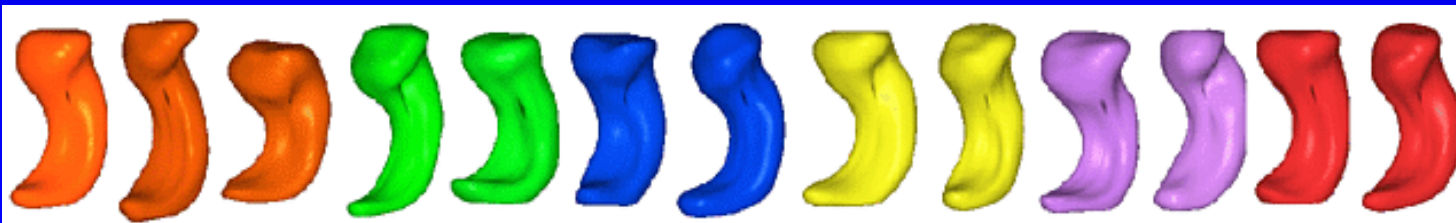
Stephen M. Pizer, Kenan Professor

Medical Image Display & Analysis Group

University of North Carolina

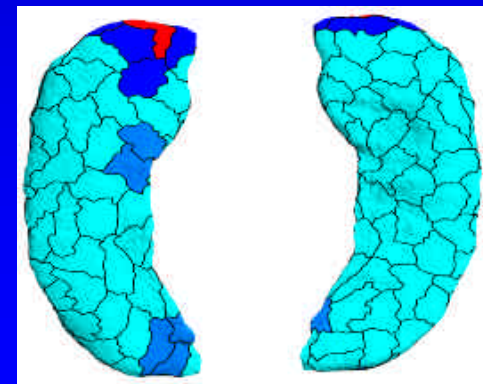
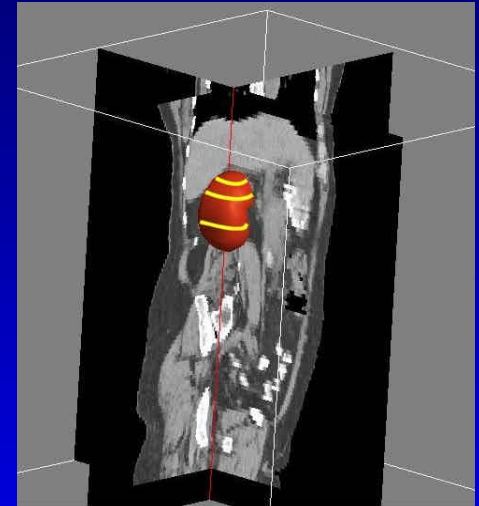
website: [midag.cs.unc.edu](http://midag.cs.unc.edu)

Co-authors: P. Thomas Fletcher, Conglin Lu,  
Sarang Joshi, Erik Dam, Roland Pilgram, Qiong Han



# Need Probability $p(\underline{\mathbf{z}})$ of Objects in Populations with $\underline{\mathbf{z}} = \text{Deformation}(\text{Mean})$

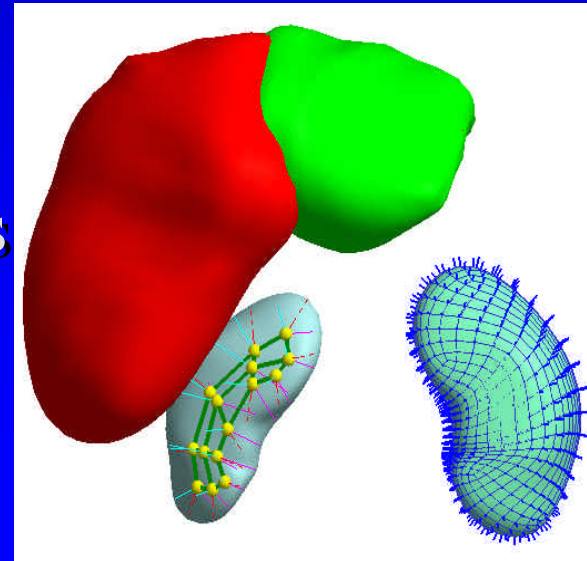
- Let  $\underline{\mathbf{z}}$  be transformation from mean, localized into object sections
- Uses for  $p(\underline{\mathbf{z}})$ 
  - Sampling  $p(\underline{\mathbf{z}})$  to communicate anatomic variability in atlases
  - Log prior in posterior optimizing deformable model segmentation
    - Optimize  $\log p(\underline{\mathbf{z}}) + \log p(\underline{\mathbf{I}}|\underline{\mathbf{z}})$
  - Compare two populations
    - Medical science: **localities where**  $p(\underline{\mathbf{z}}|\text{healthy})$  &  $p(\underline{\mathbf{z}}|\text{diseased})$  differ
    - Diagnostic: Is particular patient's geometry diseased?  $p(\underline{\mathbf{z}}|\text{healthy}, \underline{\mathbf{I}})$  vs.  $p(\underline{\mathbf{z}}|\text{diseased}, \underline{\mathbf{I}})$



# Needs of Probability Representation

## $p(\mathbf{z})$

- Accurate estimation of  $p(\mathbf{z})$  using limited samples
  - Positional correspondence
  - Multiple scale levels (beat HDLSS), incl. obj. ensembles
- Rich and intuitive characterization of geometric effects
  - Local translations, but also local rotations, magnifications
- Null probabilities for geometrically illegal objects
  - Statistics on geometric transformations
- Localization
  - Residues in multiscale framework



# Represent the Egg, not the Eggshell

## ➤ The eggshell: object boundary primitives

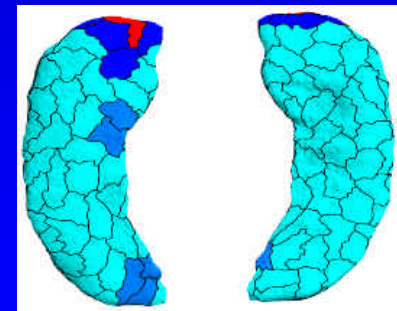
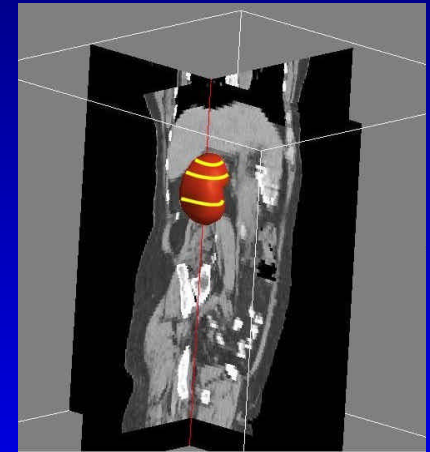
- B-reps

- Transformations of primitives: local displacement

## ➤ The egg: object interior primitives

- M-reps

- Transformations of primitives: local displacement, local bending & twisting (rotations), local swelling/contraction



# Geometric power of medial atom as basis for describing a geometric transformation

Medial atom parameters carry position, metric, two orientations

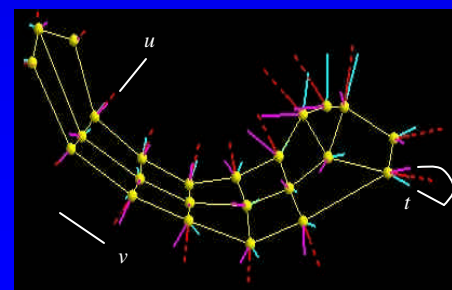
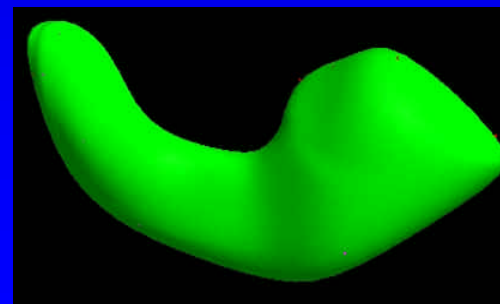
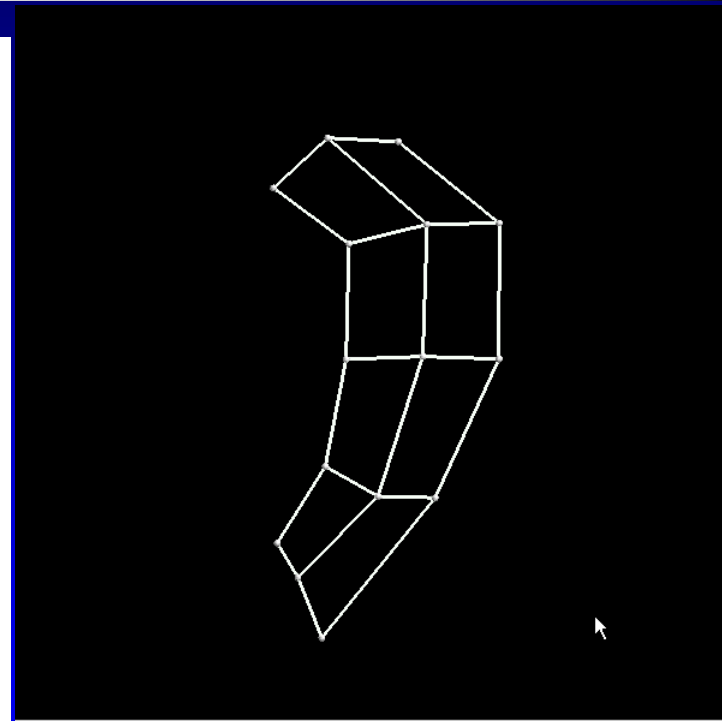
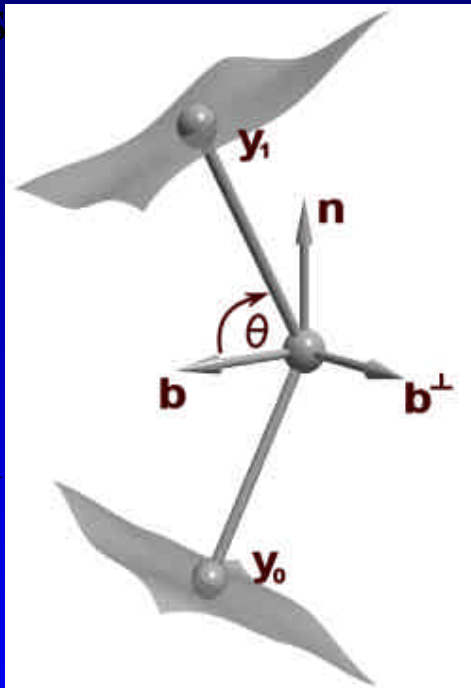
$$T \in \mathbb{R}^3 \times \mathbb{R}^+ \times S^2 \times S^2$$

From base atom:  
 Hub translation  $\times$   
 Spoke magnification in common  $\times$   
 spoke<sub>1</sub> rotation  $\times$   
 spoke<sub>2</sub> rotation

$$S^2 \equiv SO(3)/SO(2)$$

M-rep is n-tuple of medial atoms

$T^n$ , n local T's, a symmetric space



# Medial atom or m-rep as a geometric transformation, a point on a symmetric space

➤  $T \in \mathcal{R}^3 \times \mathcal{R}^+ \times S^2 \times S^2$

➤ Some m-rep parameters are not linear

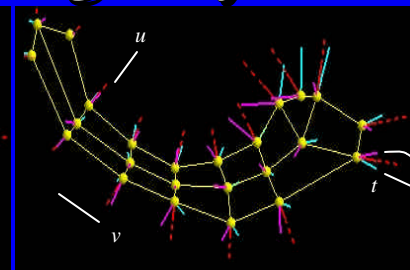
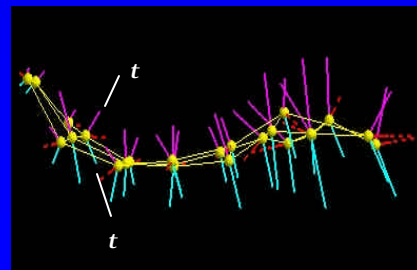
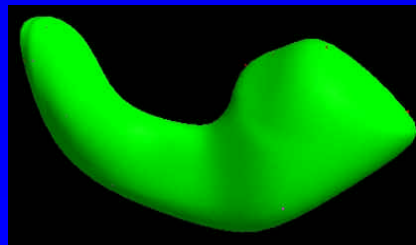
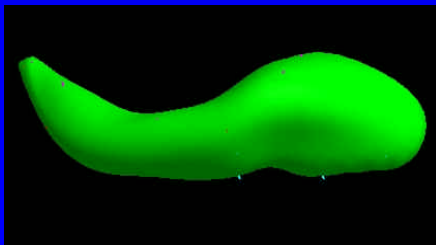
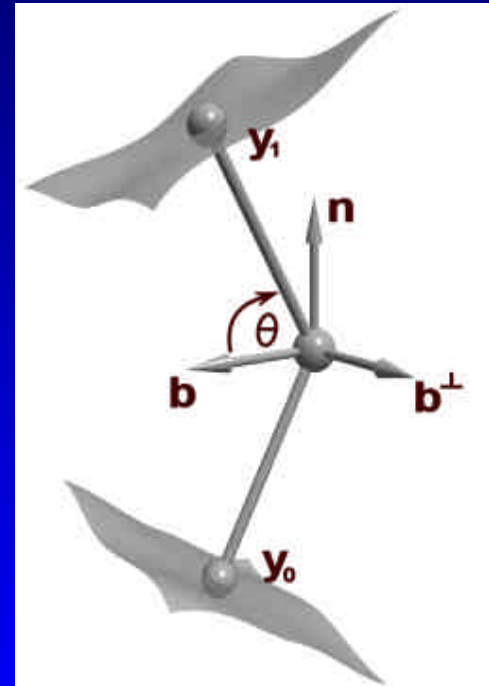
➤ Rotations

➤ Magnifications

➤ High-dimensional, curved space:

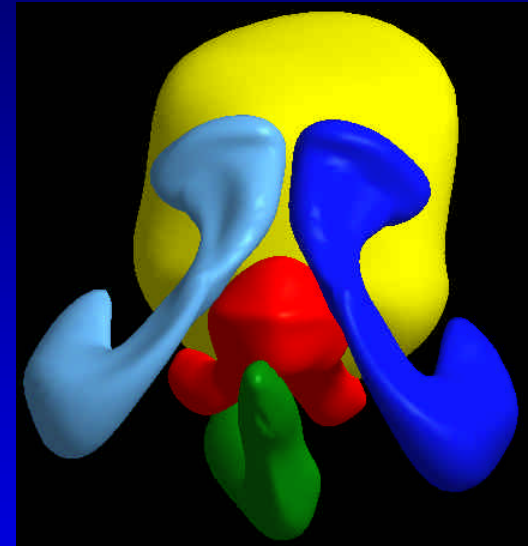
➤ Quotients of Lie groups

➤ Standard linear statistics do not achieve legality

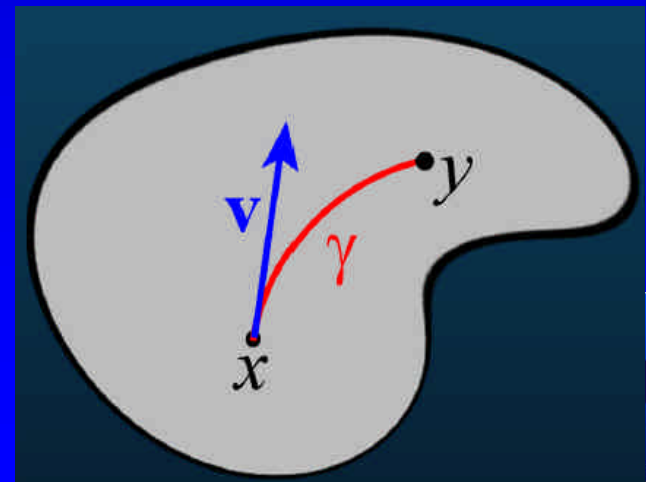
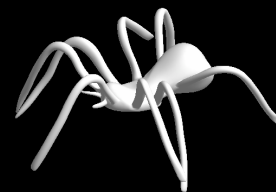
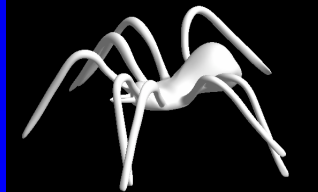
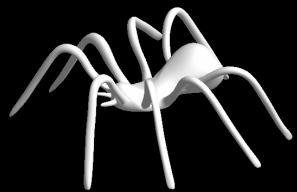


# M-reps as Points in Curved Space, Geodesics

- Deformation as **paths** of local rotation, swelling, & displacement
  - i.e., on symmetric space's surface
  - Geodesics: minimum energy
    - Geometric
    - Physical
- E.g., tweening from key frames in animation



A to B to A to C to A



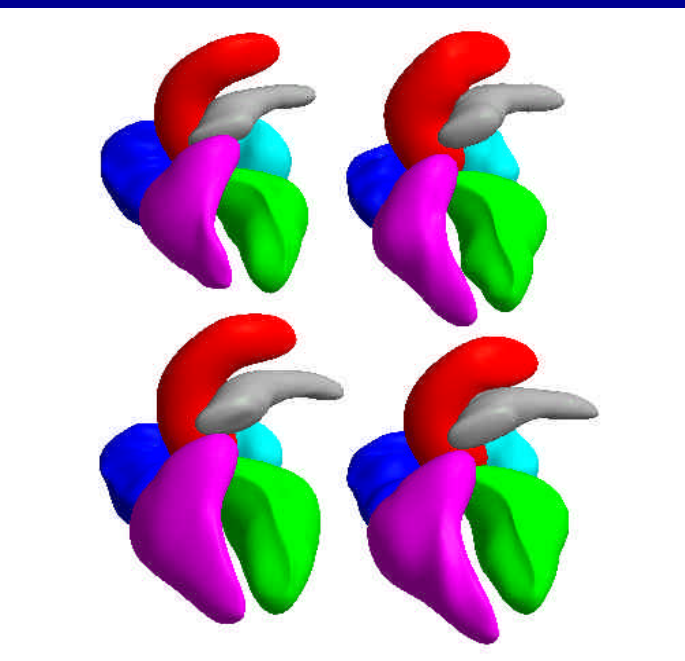
# Animation - Demo





# Principal Geodesic Analysis

## [Fletcher] - go to Fletcher presentation

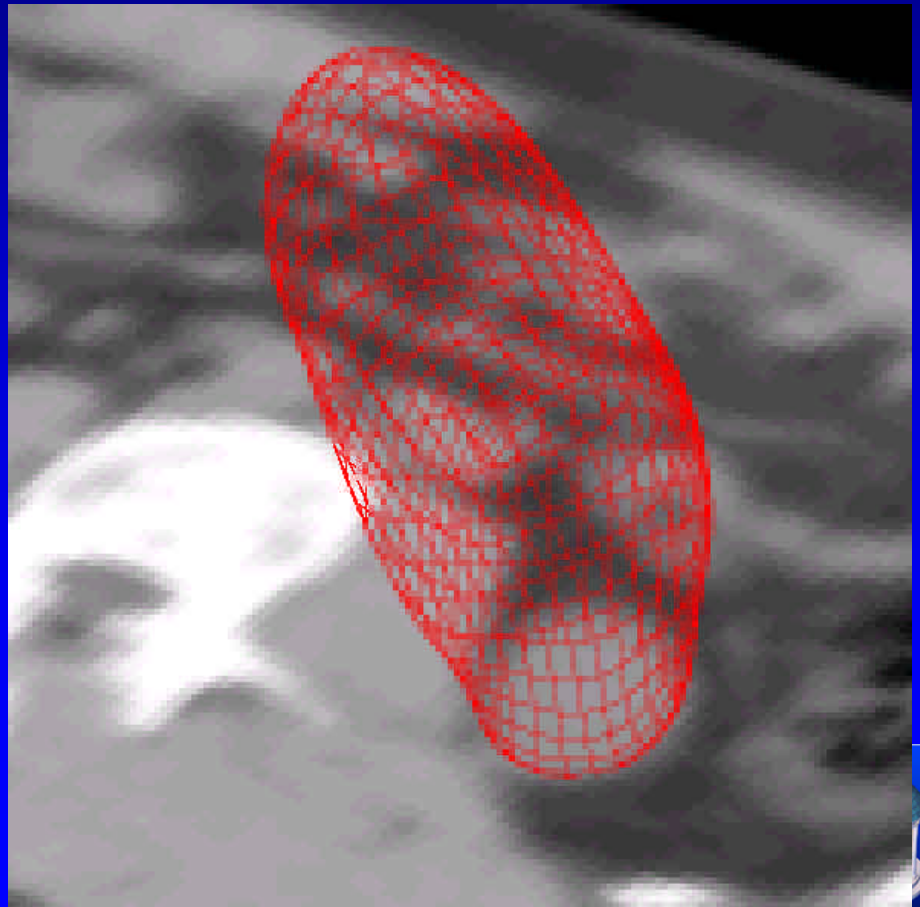
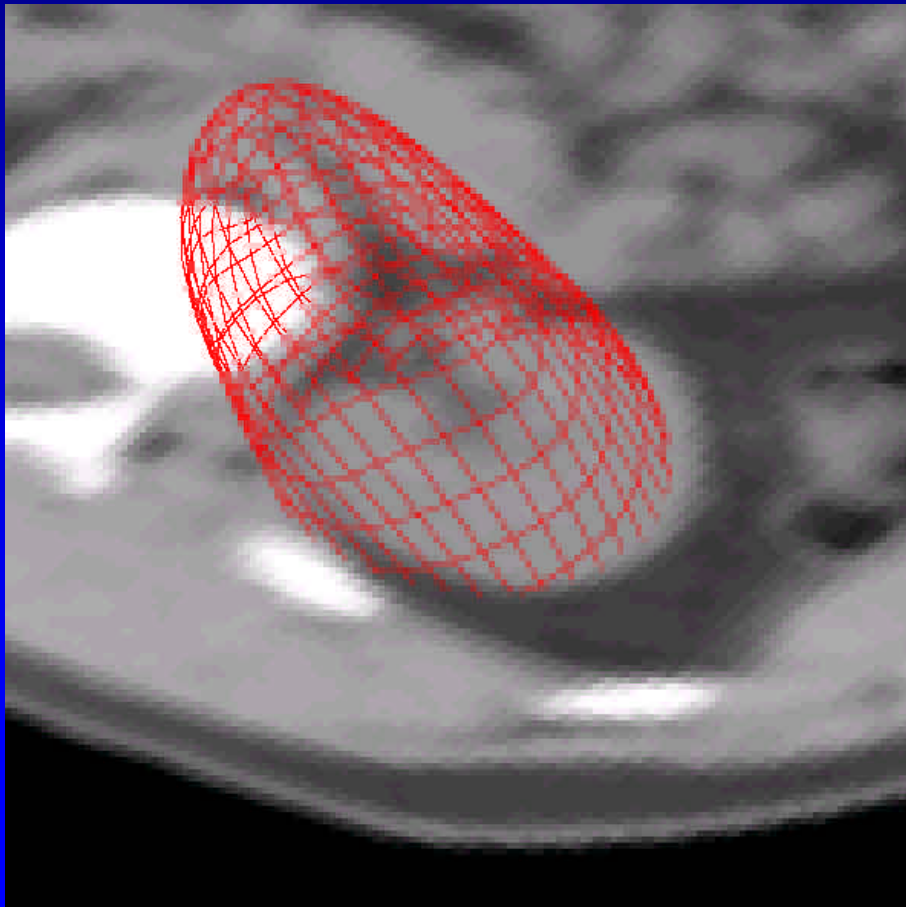


Example: 4 heart models, each made up of 7 objects [Pilgram]

- Calculated via tangent hyperplanes through mean and “exponential” & “logarithmic maps”
- Demo for object ensemble, for object

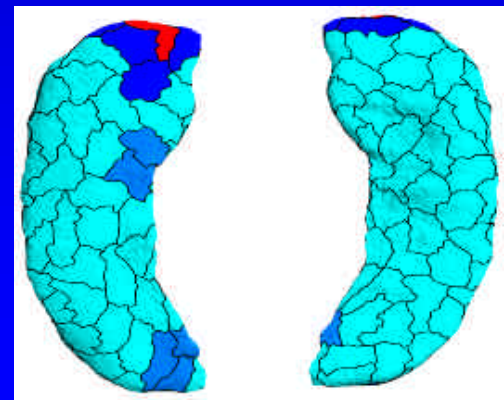
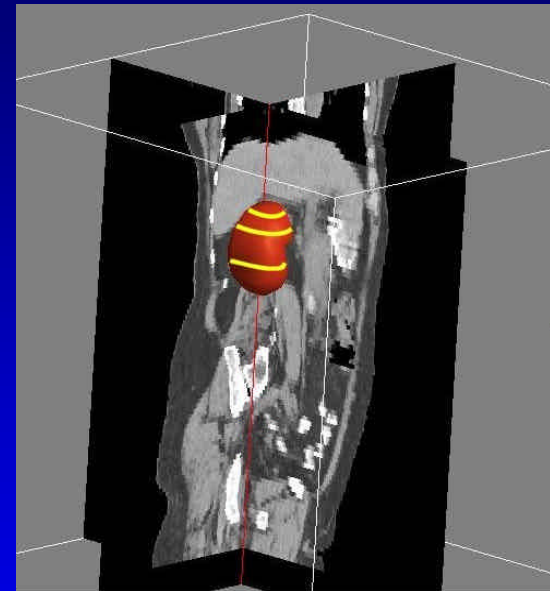
# Segmentation of a Kidney by Deformable M-reps (~10 sec.)

Optimize  $p(\underline{\mathbf{z}}|\underline{\mathbf{I}})$ , thus  $\log p(\underline{\mathbf{z}}) + \log p(\underline{\mathbf{I}}|\underline{\mathbf{z}})$



# Efficiency is Critical for Objects

- Typically more than  $10^4$  primitives
- Time efficiency
  - Optimizing probability
  - Simulated physical behavior (e.g., motion)
- Statistical efficiency: number of training samples needed
  - Inter-sample spatial correspondence
  - Rich local geometric transformations
- Both efficiencies require multiple scales (levels of locality)
  - $O(N)$  not  $O(N^2)$



# Scale Situations in Various Sampled Geometric Analysis Approaches

Global coef for each level of detail

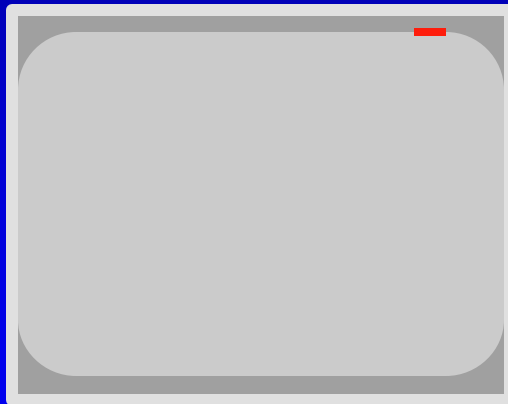
Examples: boundary spherical harmonics, global principal components

Multidetail feature Detail residues

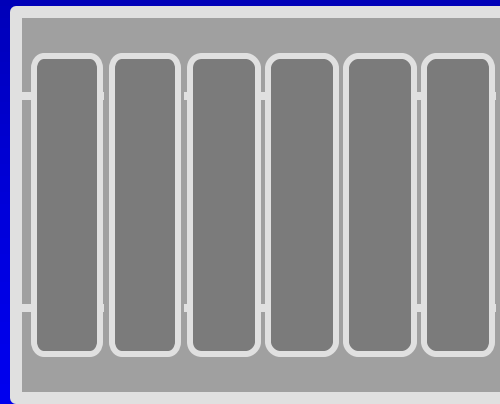
boundary points, dense position displacements, medial atoms

m-rep object hierarchy, wavelets

Level of Detail



Location

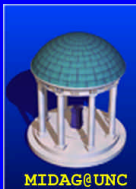
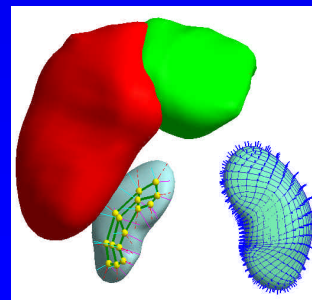


Location



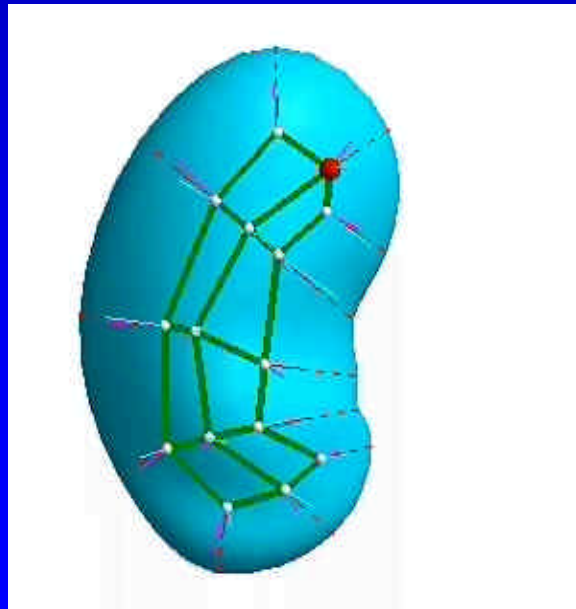
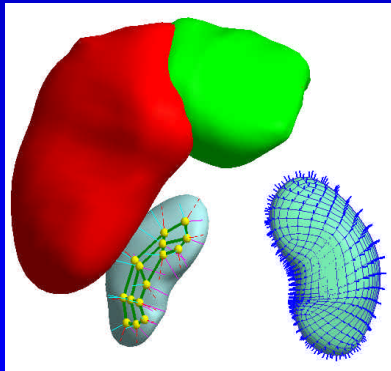
Fine

Coarse

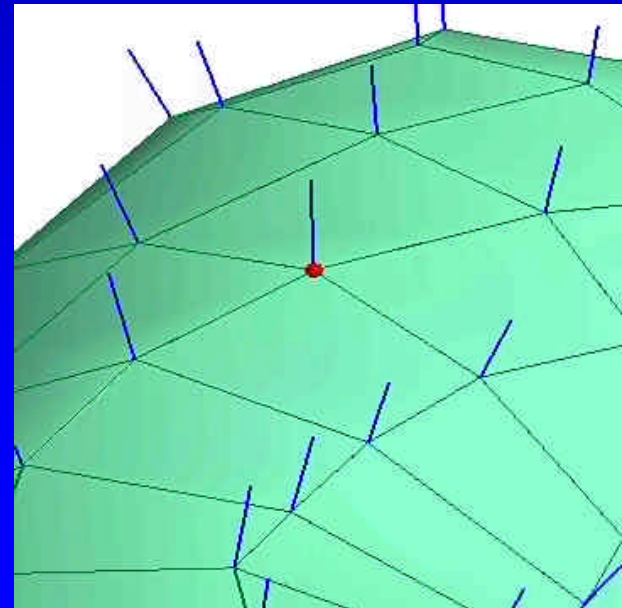


# Residues at the Object, Atom and Boundary Scale Levels

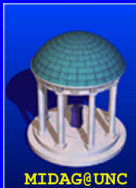
- Object transformations are transformations of all the object's atoms taken as a group
- Atom residues are  $\in \mathcal{R}^3 \times \mathcal{R}^+ \times S^2 \times S^2$
- Boundary residues (part of m-rep) are normalized displacements from medially implied boundary positions



atom residue



boundary residue



# Markov Random Field Models for Primitives with Neighbor Relations

## ➤ Markov assumptions

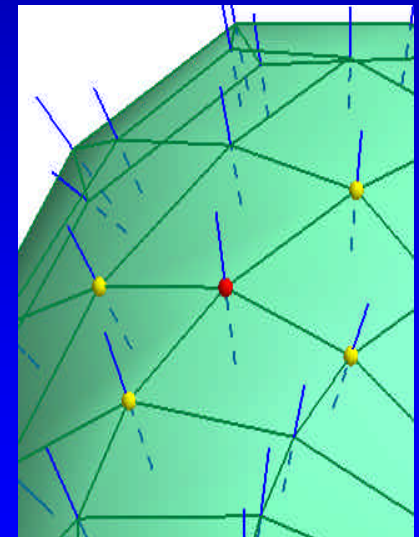
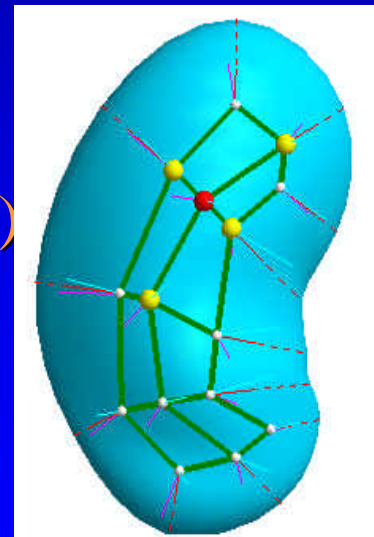
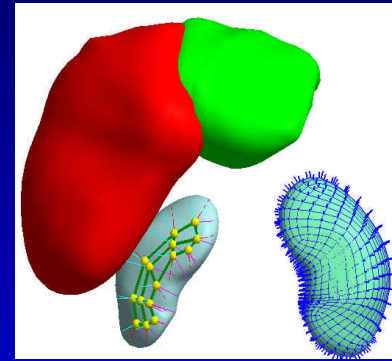
- Inter-scale residues ( $\Delta T^m = T^m - T^{m-1}$  models the residue at scale level  $m$ )

$$\text{Prob}(T^m | \{T^s, s < m\}) = \text{Prob}(T^m | T^{m-1}) = \text{Prob}(\Delta T^m)$$

- At entity  $i$ , model  $\text{Prob}(\Delta T^m_i)$ , where

$$T^m_i = \text{Pred}(T^m_{\{N(i,j)\}}) + \Delta T^m_{(i,j)}$$

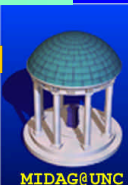
- Both have moderate dimension



atom level

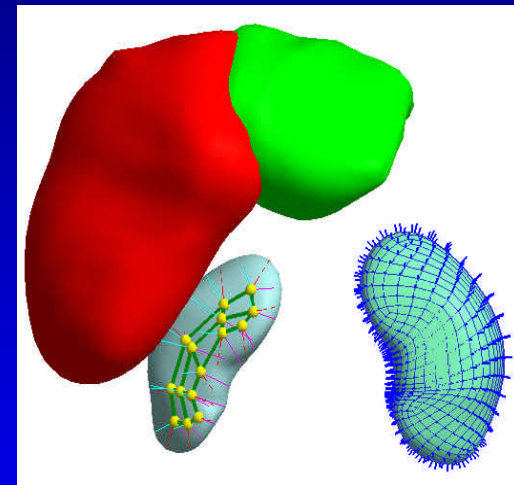
boundary level

quad-mesh neighbor relations



# Deformation Parameters by Scale Level for M-reps

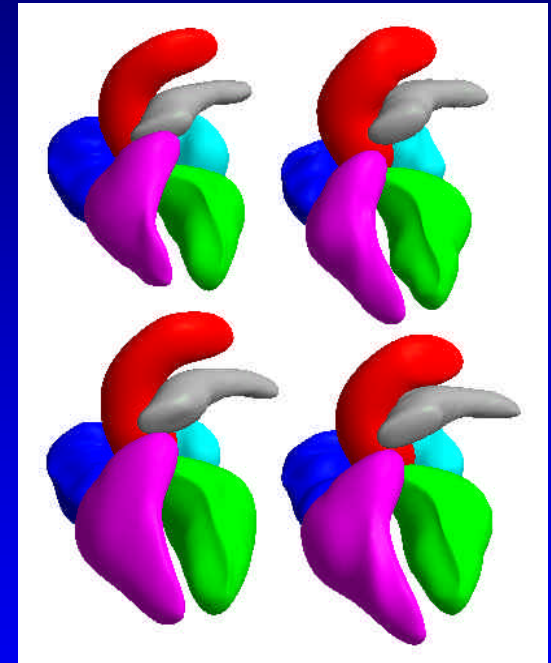
Object ensemble	Sim. Transf & PGC coef's
Object	Sim. Transf & PGC coef's
Main figure	Sim. Transf & PGC coef's
Subfigure	Sim. Transf. in host (u,v,t) & PGC coef's
Medial atom	Atom parameters
Boundary vertex	Displacement along normal (medial atom spoke)



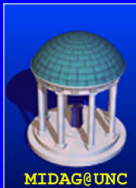
# Training M-rep Probabilities

## [Fletcher, Han, Dam]

- Start with
  - N binary images of segmentations
  - # of figures and medial atom grid size of each figure
- Fit m-reps to images
- Regularize m-reps and refit
- Align using Lie group distances
- Compute mean and principal geodesics at top scale
- Fit PGCs to cases and refit residues



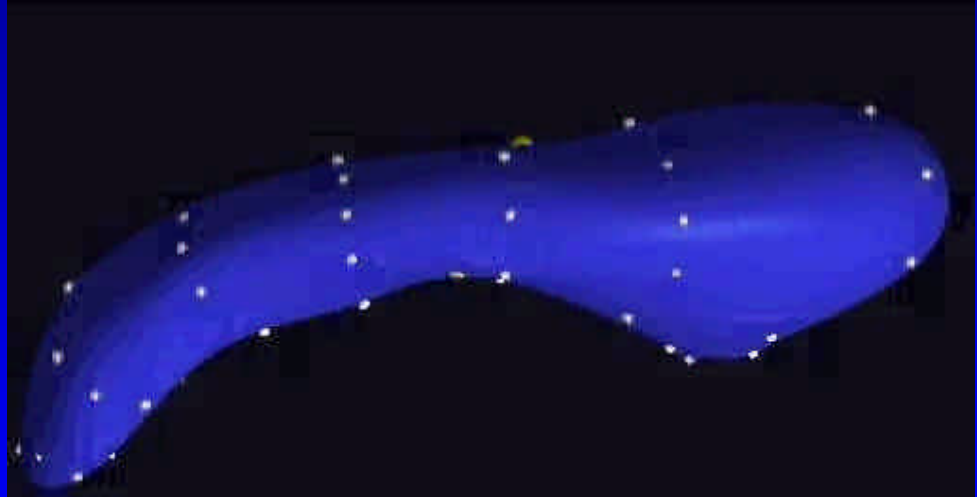
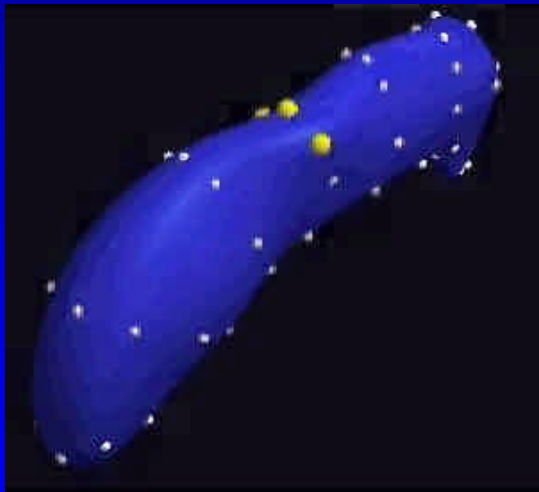
A seven-object heart  
(in future 3 multfigure objects:  
pericardium, right, left)





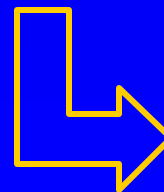
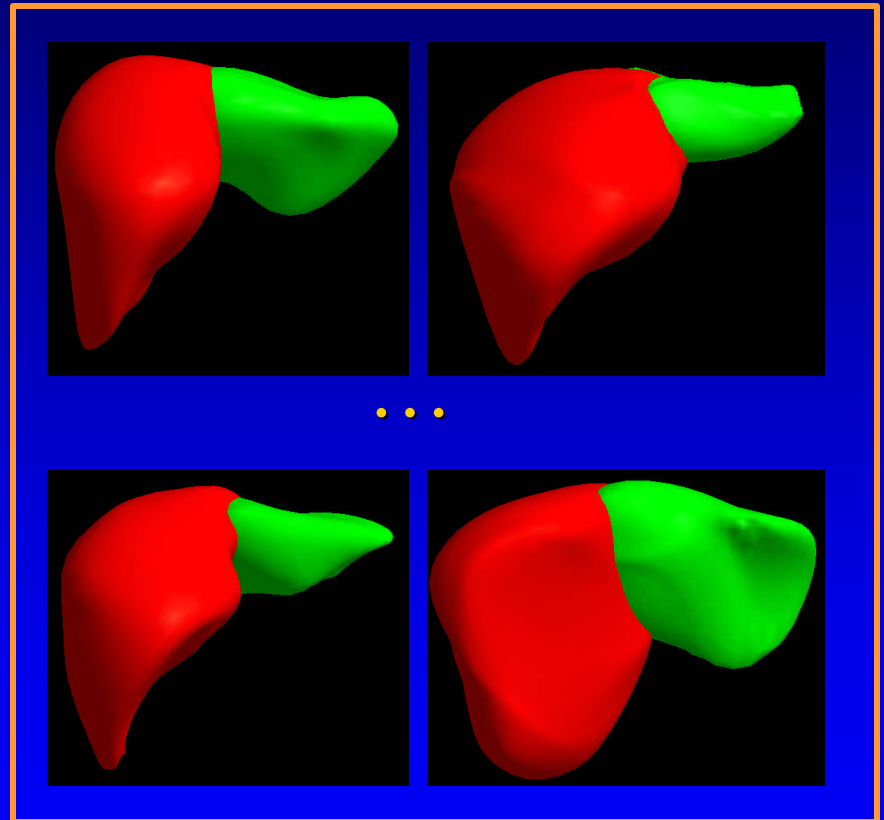
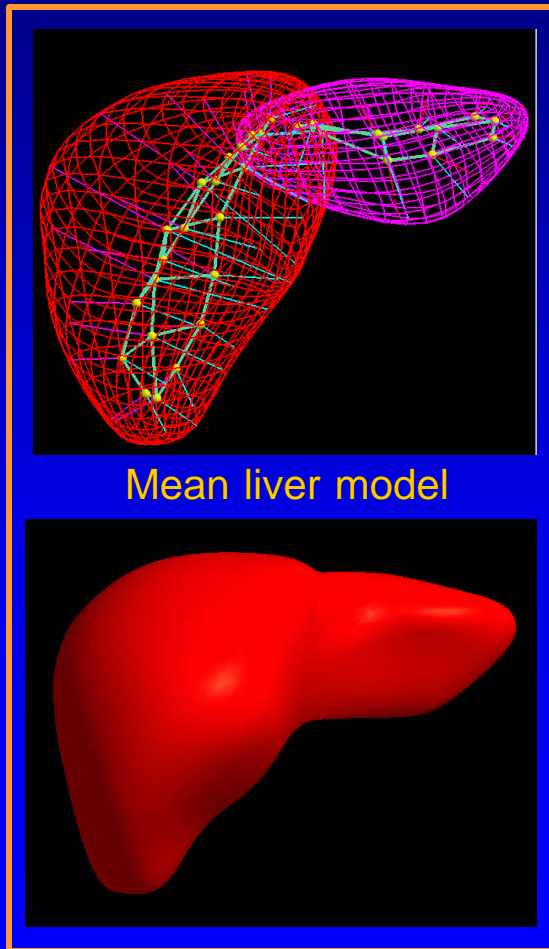
# Hippocampus medial atom (residue) statistics ([Lu])

- Global over hippocampus vs. by locality
- Residue
  - From next larger scale (vs. from neighbors)



- Demo

# Mean and PGCs for Liver [Lu]

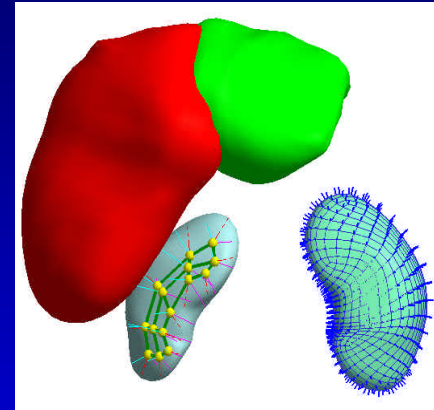


Principal modes

# Our Approach to Statistical Analysis of Object Geometry

- Use m-reps: medial representation together with boundary displacement
  - primitive transformations decomposable into translations, rotations, magnifications

[Pizer, IJCV 2003 and Joshi, TMI 2002]



- Build Markov random field models for residues
  - inter-scale residues between scale levels
  - intra-scale residues among neighboring primitives at each scale level

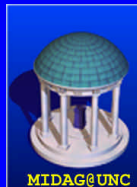
[Lu, Scale Space 2003]

- Residue models of geometric transformations and the metrics on them are defined and analyzed via geodesics on symmetric space [Fletcher, CVPR & IPMI 2003]



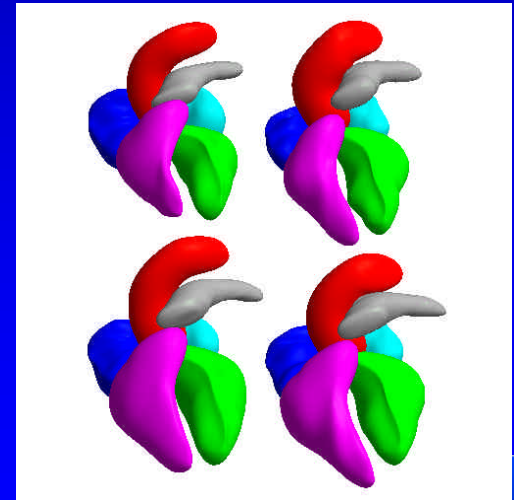
# Other Applications of Lie Group and Lie Group Quotient Statistics

- Statistics of tensors (from DTI) [Fletcher]
- Statistics of neural fiber bundle geometry [Gerig]
- Statistics of joystick gestures [Feasel]
- Extension to diffeomorphism group
  - Incl. as small scale residue within m-reps
- Any statistical pattern recognition method, previously on linear space
  - Discrimination
    - Fisher linear discriminant
    - Support vector machines
    - Kernel methods
    - Feature selection with locality [Yushkevich]
  - Clustering



# Future Work

- MRF models at other scale levels
  - ⇒ complete probabilistic model on all scale levels
- Test adequacy of close-neighbor Markov assumption
- Connect image and geometric scale spaces
- Localized feature selection
  - On 3D medial
  - With multiscale residues
- Statistics of multiple multigure objects



For background to this talk see tutorial at website:

[midag.cs.unc.edu/projects/object-shape/tutorial/index.htm](http://midag.cs.unc.edu/projects/object-shape/tutorial/index.htm)

[http://midag.cs.unc.edu/pubs/presentations/Pizer\\_SPIE.ppt](http://midag.cs.unc.edu/pubs/presentations/Pizer_SPIE.ppt)

[http://midag.cs.unc.edu/pubs/presentations/Joshi\\_SPIE.ppt](http://midag.cs.unc.edu/pubs/presentations/Joshi_SPIE.ppt)

[http://midag.cs.unc.edu/pubs/presentations/Gerig\\_SPIE.ppt](http://midag.cs.unc.edu/pubs/presentations/Gerig_SPIE.ppt)

or papers at

[midag.cs.unc.edu](http://midag.cs.unc.edu) bibliography

