

Topo-Geometric Modeling of 3D Objects

Hamid Krim,
ECE Dept., NCSU,
Raleigh, NC.

Acknowledgements:

D. Aouada, S. Baloch, A. Ben Hamza, S. Feng

D. Dreisigmeyer, I. Kogan

*AFOSR and ONR

Outline

- 3D Shape Modeling
 - Topological modeling
 - Geometric modeling
- Geometric model matching
 - Integral Invariants
 - Differential Invariants
 - Riemannian metric
- Conclusion

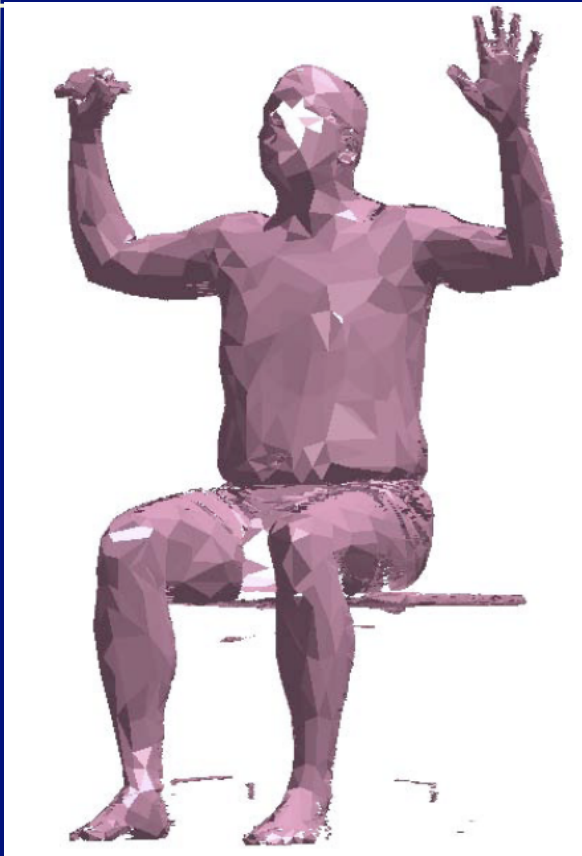
Motivation for 3D shapes

- Generalized framework for classification and recognition
- Biomedical imaging (surgery assistance...)
- Object compression for storage/retrieval
- CAD applications, Art archival, terrain modeling

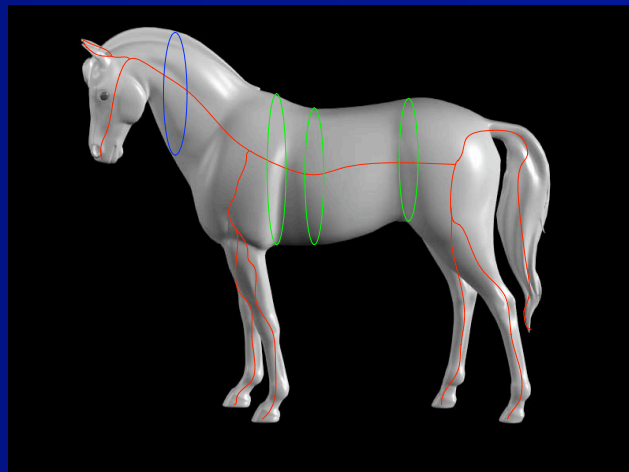
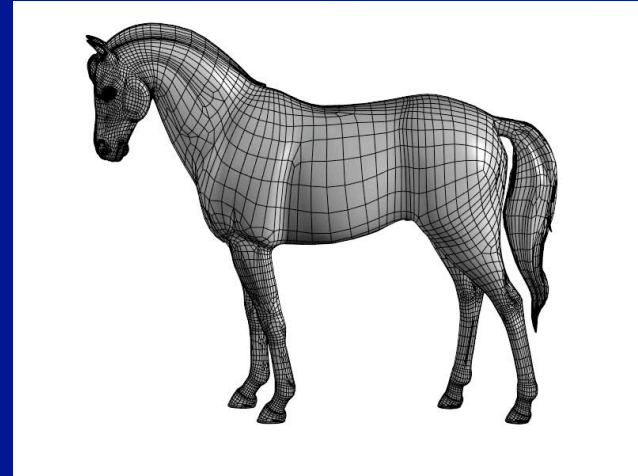
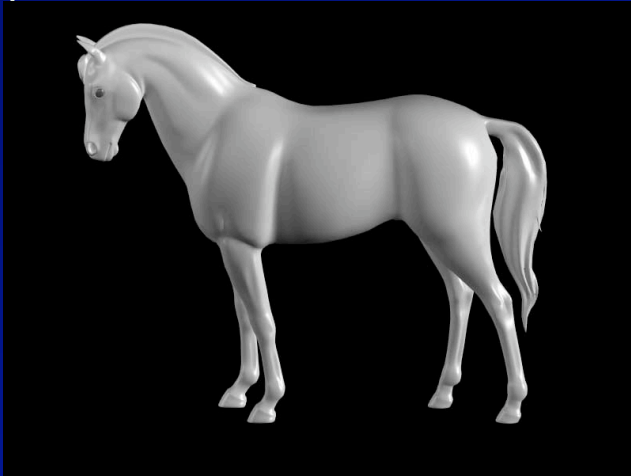
Other 2D/3D Representations

- Shock graphs
[Kimia-Tannenbaum-Zucker, Siddiqi-Zucker....]
- Medial Axis
[Blum, Damon, Giblin, Kimia, Pizer, Siddiqi]
- M-Reps [Nackman-Pizer, Fletcher....]
- Morse-Theoretic
[Shinagawa, Kunii and Kergosien, Schroeder,
Edelsbrunner, Schmidt *et al.*, Andres *et al.*,
Samir *et al.* ...]

Coordinate-Free Representation



Pictorially...



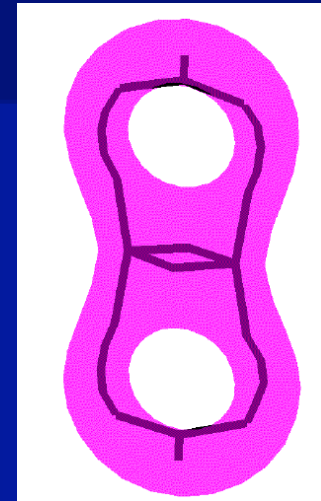
In words...

Coarse representation

- Fast and simple tool for surface comparison

Classification

Topological
or skeletal
graph

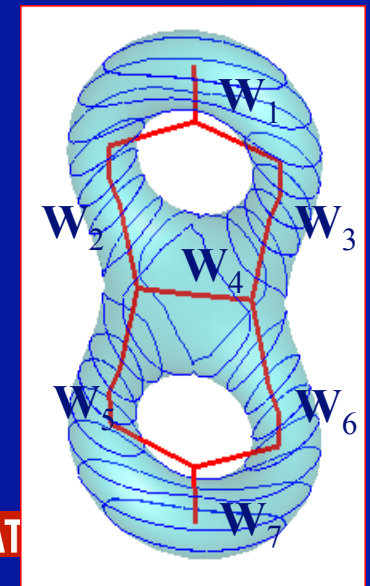


Fine representation

- Higher level of discrimination.

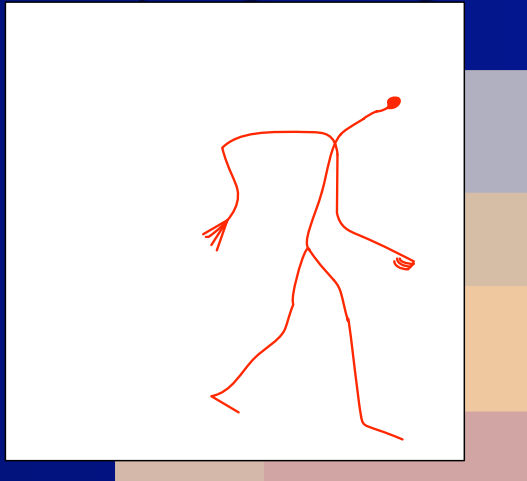
Recognition

Geometric
information



Topology

- *Goal: Represent a surface/manifold in subparts which may be glued together*



- **Information in Topology**
- **How to capture topology?**
 - **Critical** points

Morse theory

- Study of smooth and compact surfaces by exploiting a defined Morse function

Definition:

A smooth function $h : \mathcal{S} \rightarrow \mathbb{R}$ on a smooth manifold \mathcal{S} is called a *Morse* function if all of its critical points are non-degenerate.

- A critical point $p_0 = (u_0, v_0)$ is degenerate if the Hessian of $h \circ x$ is singular

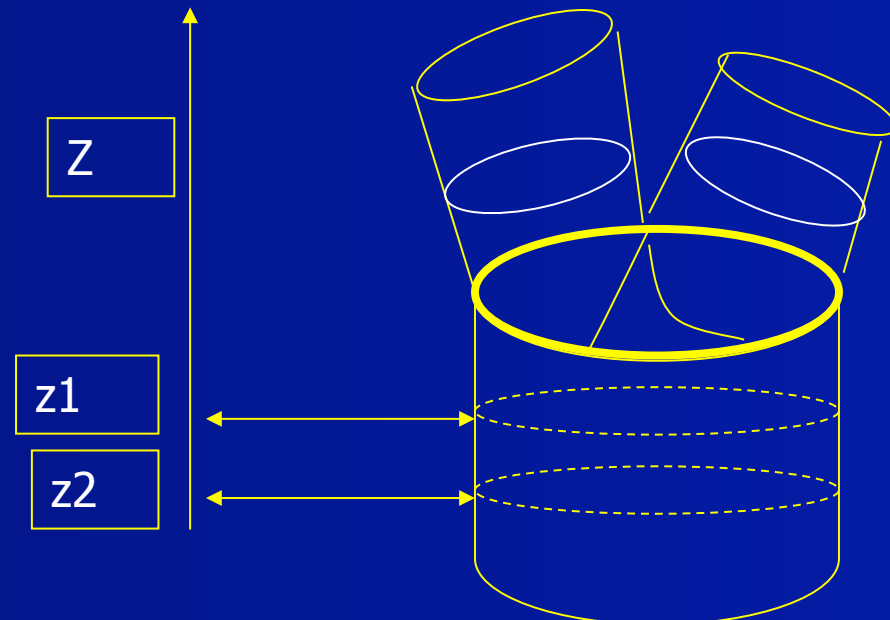
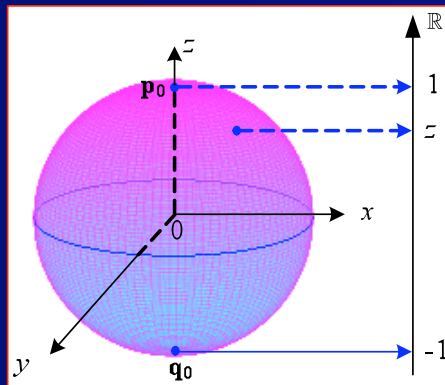
Properties:

- Critical points of a Morse function are:
 - isolated
 - landmarks on the surface
- Saddle points determine topological changes of the surface

Height Function

- A height function $h: \mathcal{M} \rightarrow \mathbb{R}$ on smooth manifold is a real valued function such that

$$h(x, y, z) = z, \forall (x, y, z) \in \mathcal{M}$$



Reeb graph



Not invariant to
•Rotation

– Reeb graph may alternatively be described as a **quotient space** \mathcal{M} / \sim where the **equivalence relation** \sim is defined as:

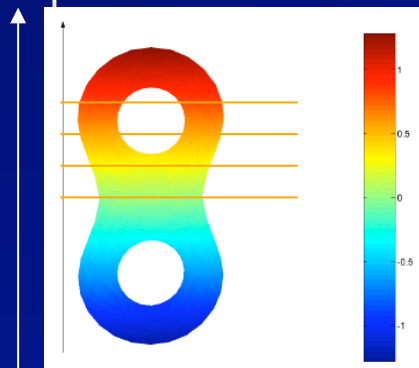
– $\mathbf{p} \sim \mathbf{q}$ iff

▪ $h(\mathbf{p}) = h(\mathbf{q})$ $\mathbf{p} \in \text{ConnComp}(\text{Levelset}(\mathbf{q})) = h^{-1}(h(\mathbf{q}))$

$$\mathcal{M} / \sim := \{[\mathbf{p}] : \mathbf{p} \in \mathcal{M}\}$$

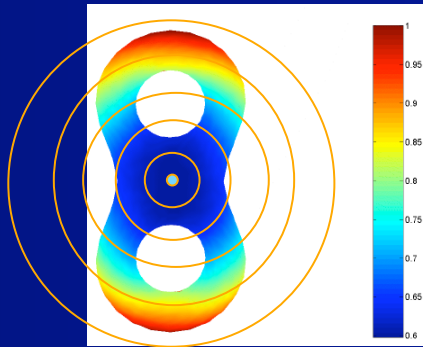
$$[\mathbf{p}] = \{\mathbf{q} \in \mathcal{M} : \mathbf{q} \sim \mathbf{p}\}$$

Choice of a Morse function



Height Function:

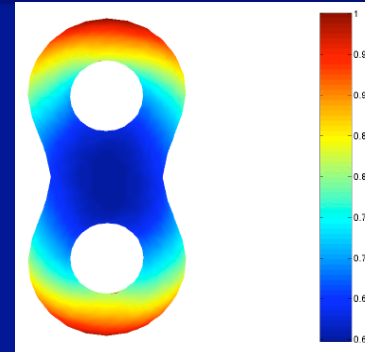
Not invariant to rotation. (*Reeb, Shinagawa et al., Edelsbrunner et al., Ben Hamza et al.*)



Spherical sampling of a surface:

Requires a reference point.

(*S. Baloch and K.*)



Geodesic Function

Completely intrinsic to a surface.

(*Hilaga et al.*)

Global Geodesic Function

- Integrated geodesic distance

$$f(\mathbf{v}) = \int_{\mathbf{p} \in S} d(\mathbf{v}, \mathbf{p}) dS.$$

- Given a 2D surface as a 3D mesh represented by $\mathbf{p}_{\{i=1, \dots, m\}}$ vertices
- $d(\cdot)$ being the geodesic distance between two vertices, GGF $g(\cdot)$ is defined as:

Global Geodesic Function *GGF* (2)

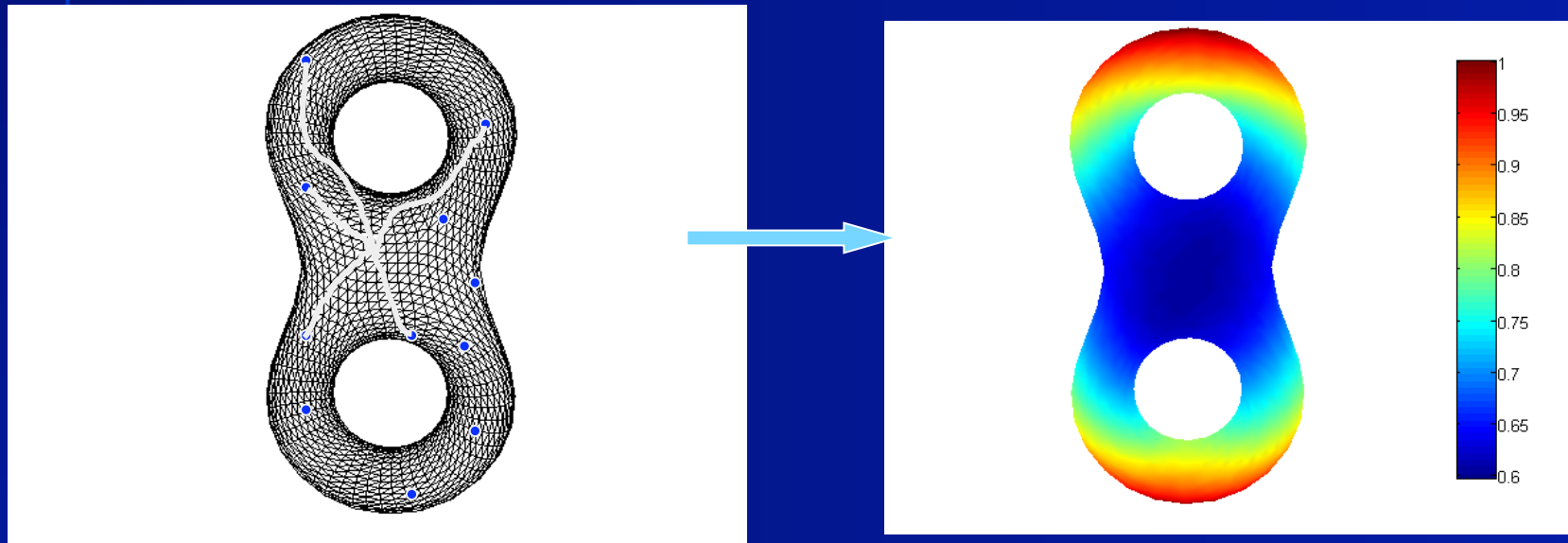
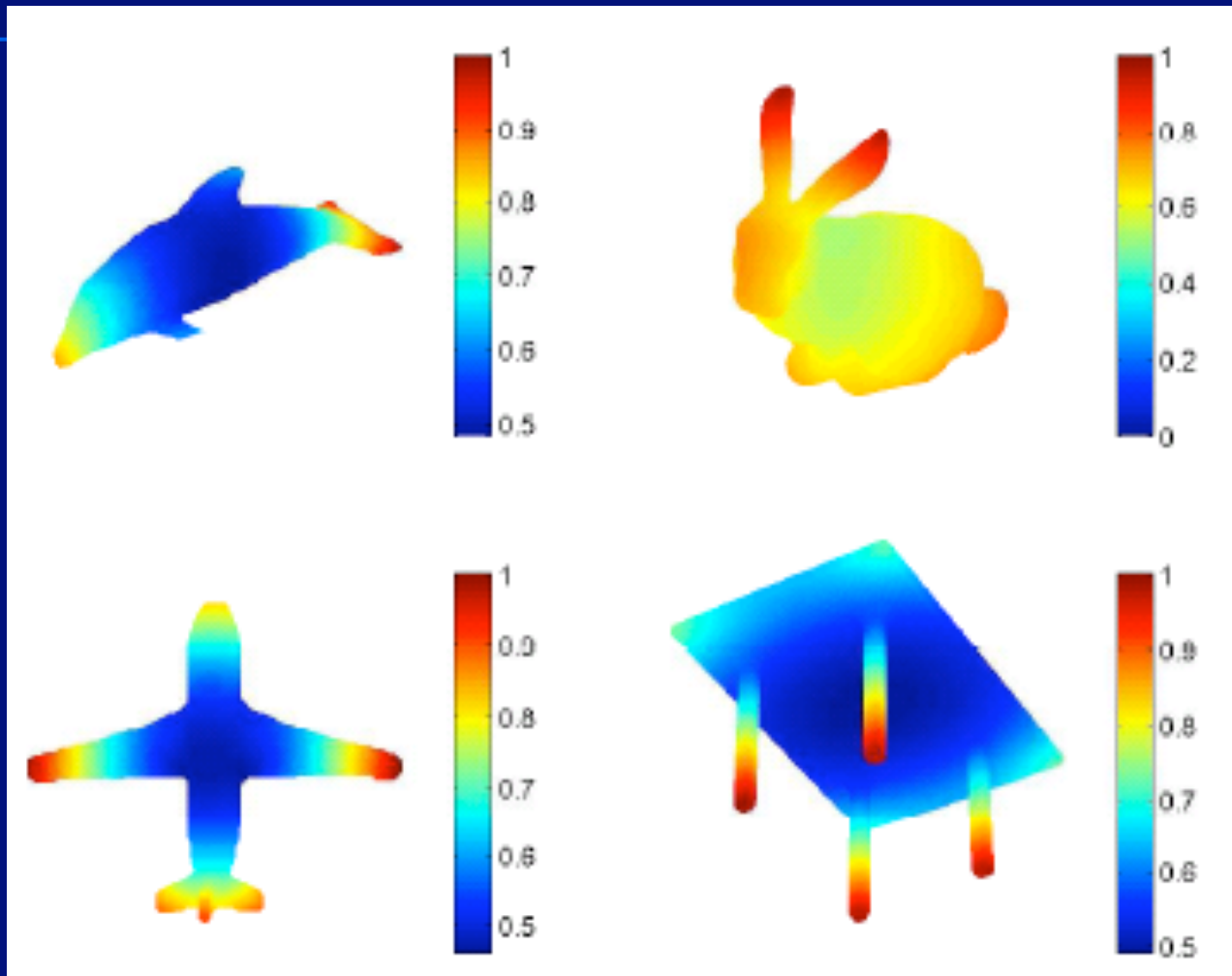
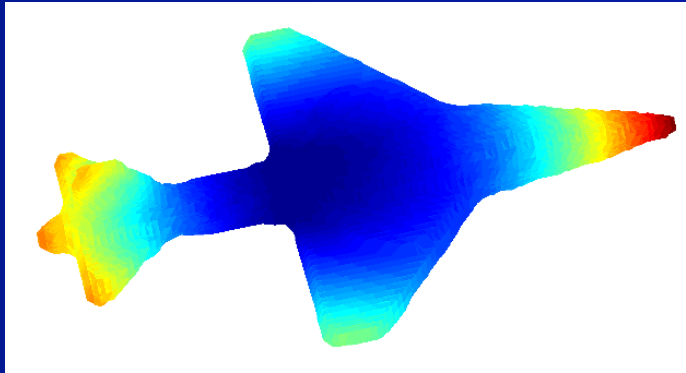
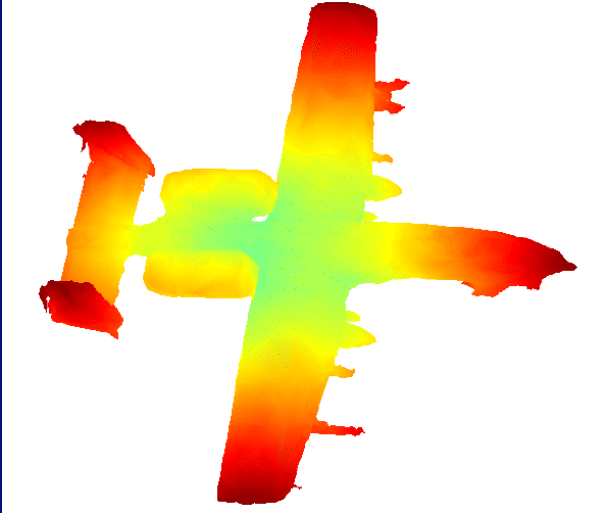
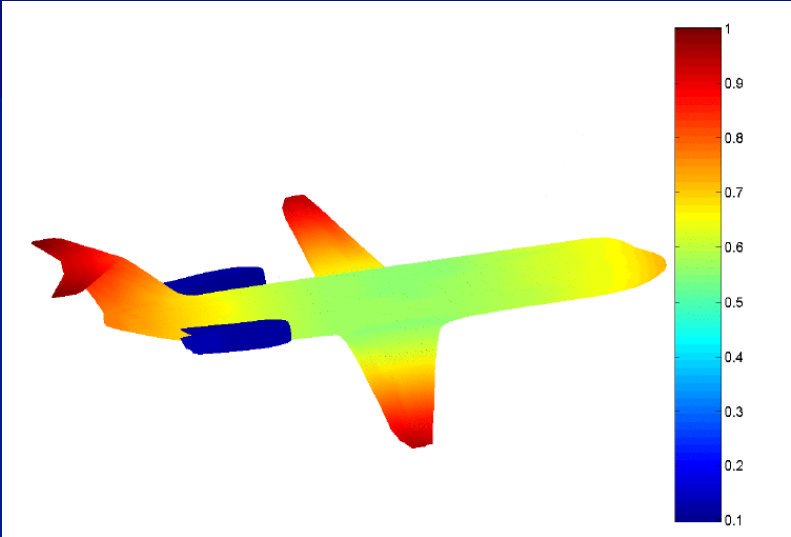
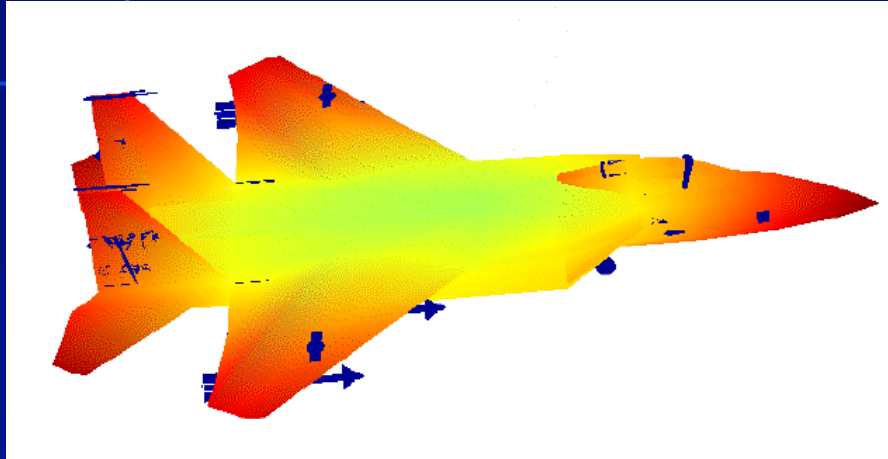


Illustration of the GGF

Global Geodesic Function GGF





VISSTA

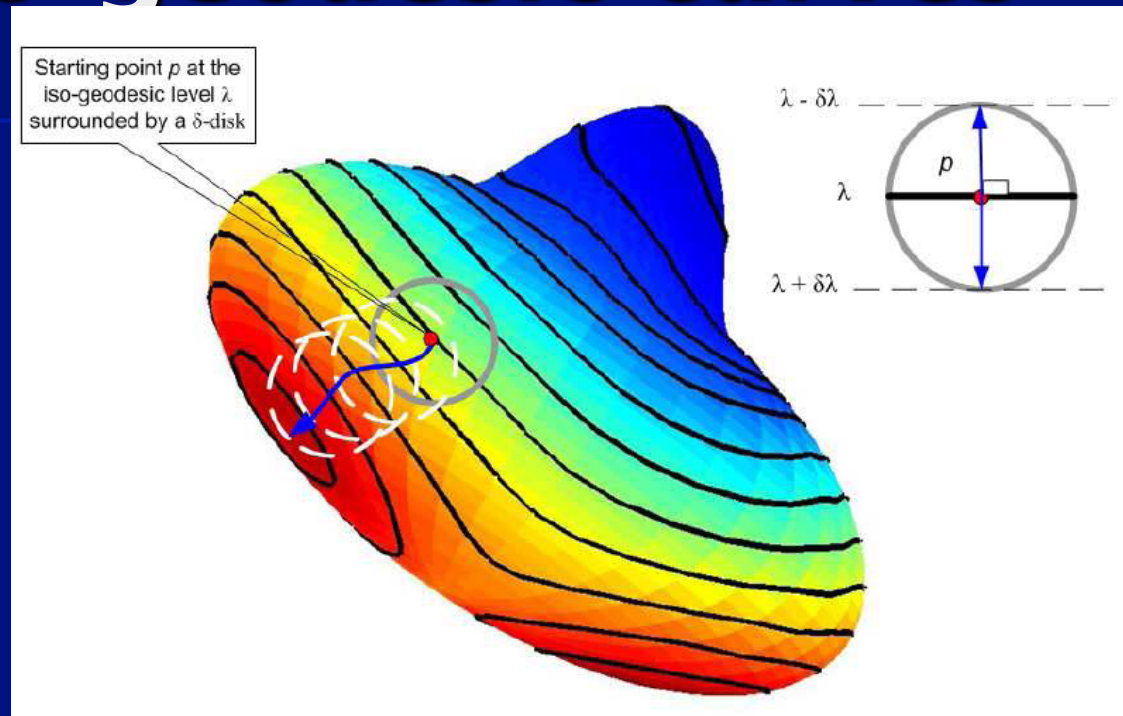
NC STATE UNIVERSITY

TG Characterization

- Morse framework provides a powerful means for gleaning topological and geometric information,
- For a surface $F(x,y,z)$ and a height function $h(f(.),.)$, we seek level sets

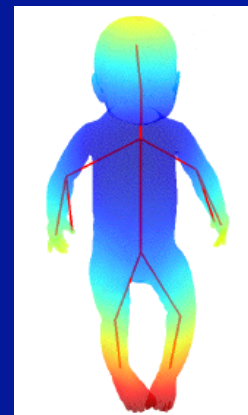
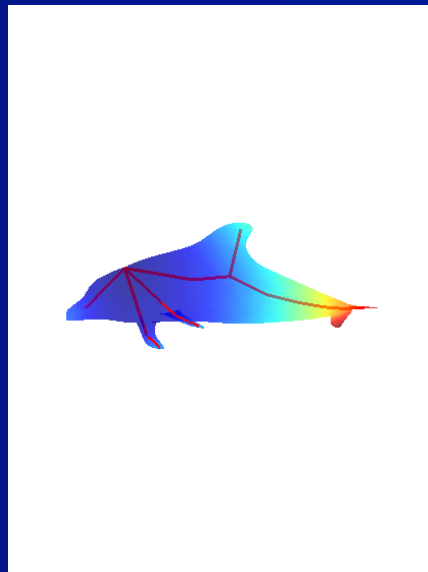
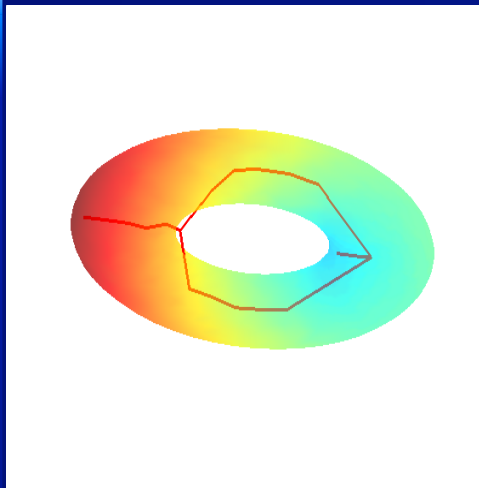
$$F(x, y, z) \cdot (h(x, y, f(x, y)) - C_t) = 0$$

Iso-geodesic curves

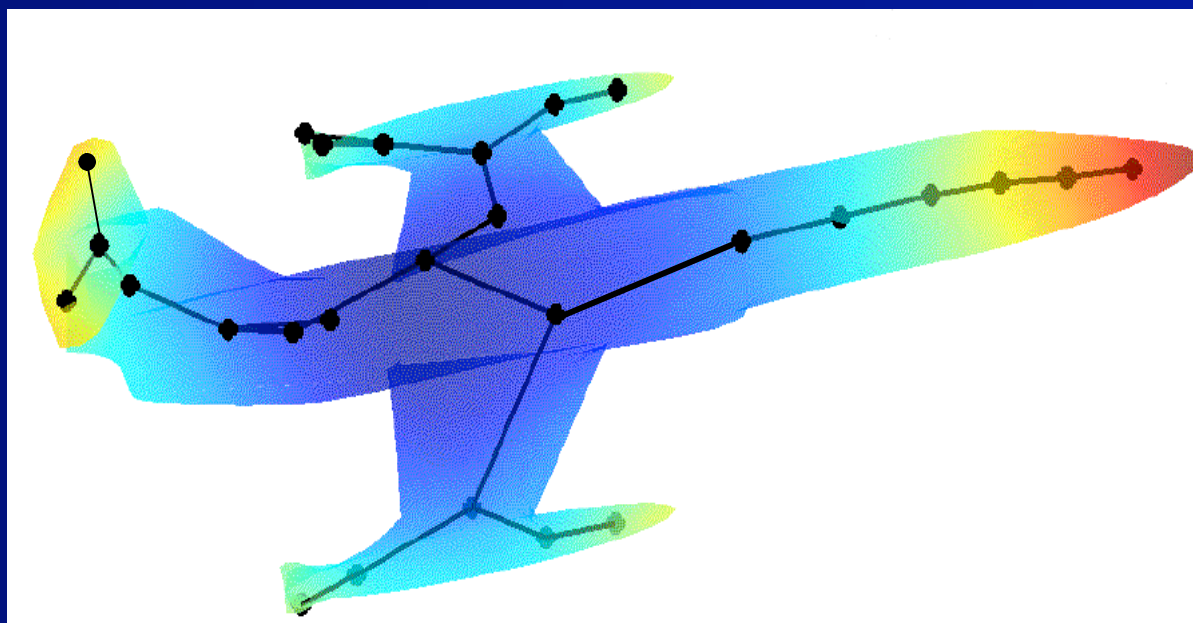


- Iso-geodesic curves are smoothly interpolated.
- The adjacency of nodes is verified through path connectedness. (with D. Aouada)

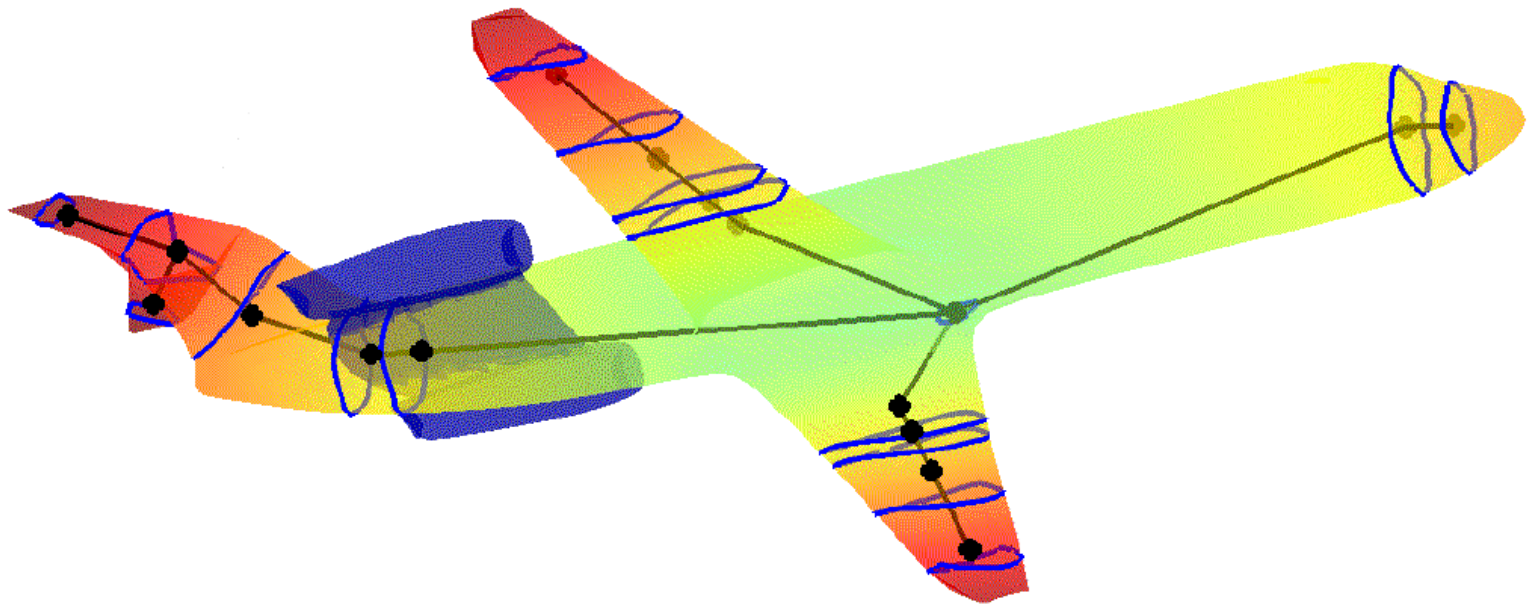
Examples of Reeb-like graphs



Topological Characterization

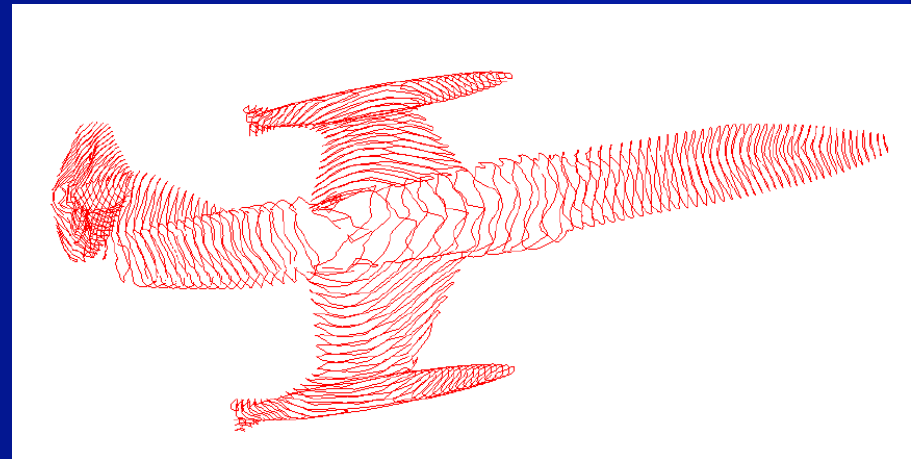
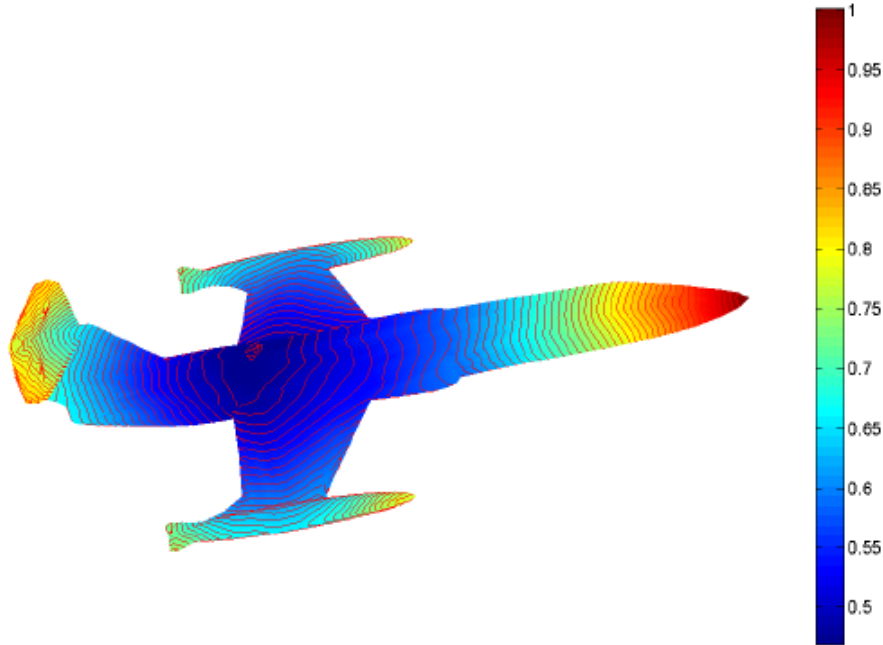


Topological Characterization (2)

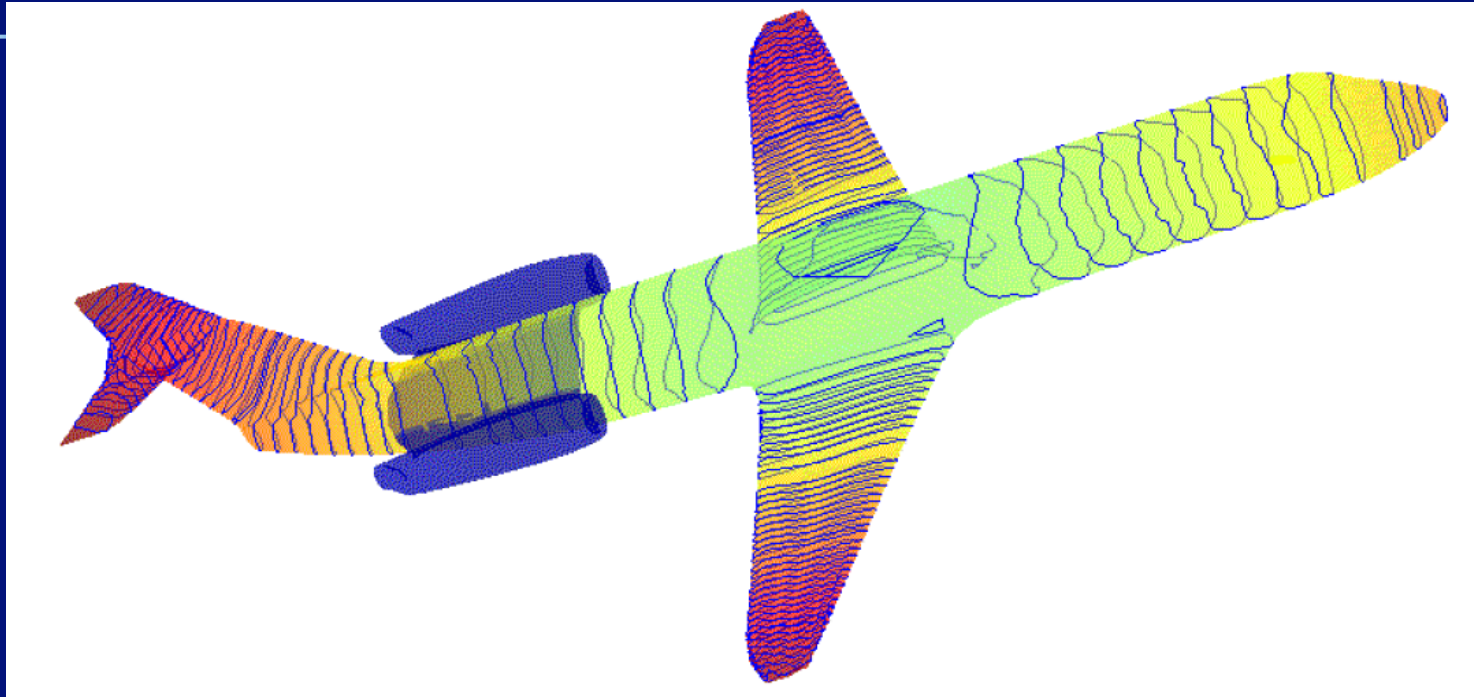


- The adjacency of nodes is verified through path connectedness.

Geometric characterization

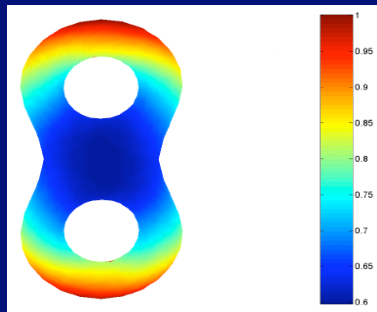


Geometric characterization (2)

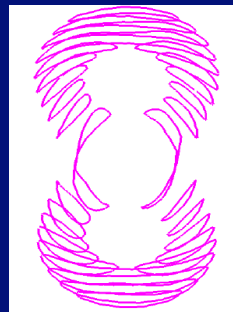


- Surface sampling.
- Iso-geodesic curves are smoothly interpolated.

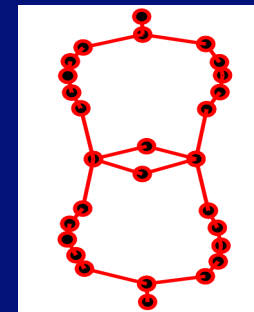
Tag modeling



GGF of the surface



Surface sampling.



Topological graph

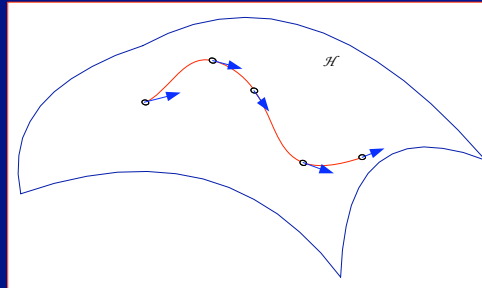


Assign a geometric weight to each edge

Iso-geodesic curves bear geometric features: Find a model to simply represent a set of these curves along each edge.

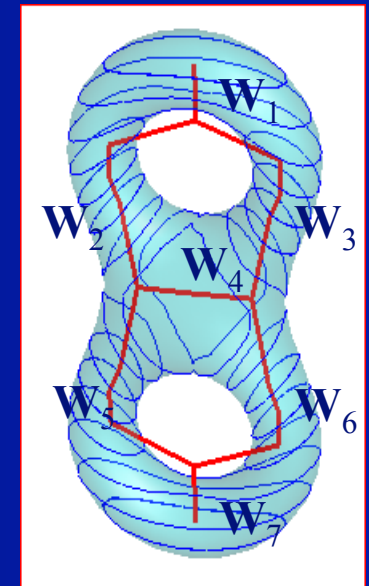
Geometric modeling (1)

- Capturing geometry



- TaG weighted skeletal graphs

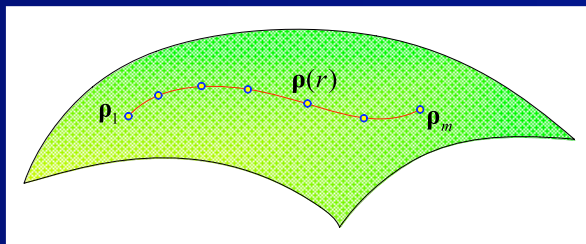
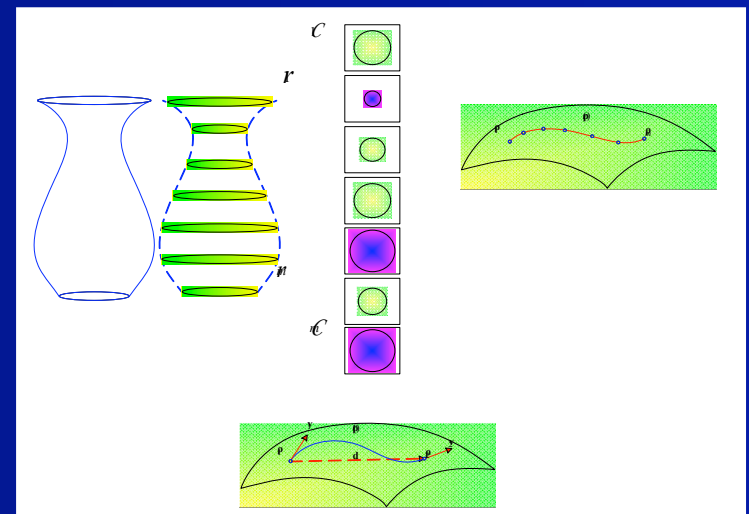
- Unique, compact and “complete” representation of 3D shape



Geometric Modeling (2)

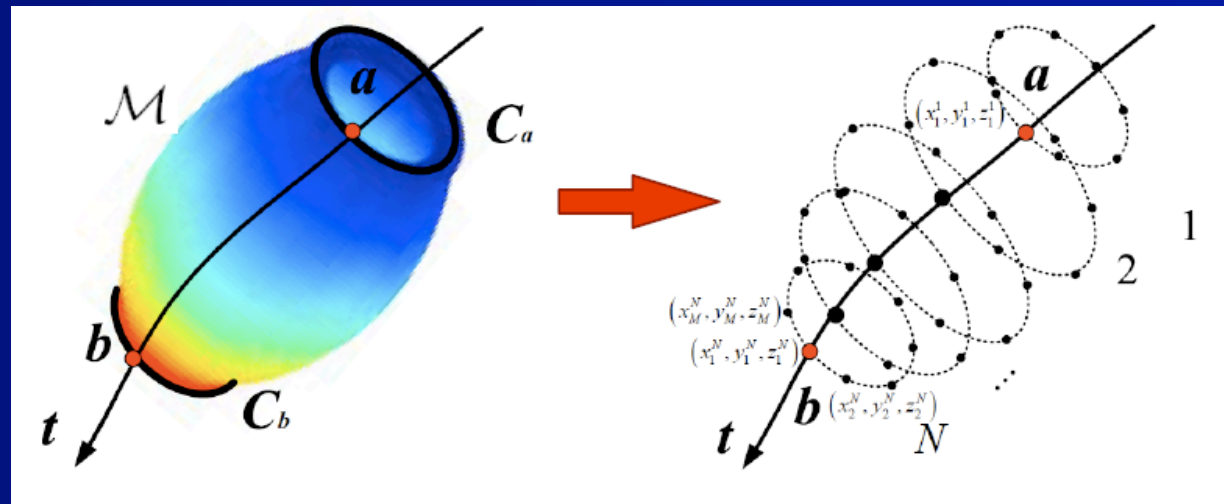
- Given m curves C_1, \dots, C_m , at levels r_1, \dots, r_m

- Find optimal trajectory $\alpha: I \rightarrow \mathbb{R}^n$



S.H. Baloch, H.K., W. Mio, and A. Srivastava,
"3D curve interpolation and object
reconstruction", *IEEE ICIP*, 2005.

In Summary



$$\mathcal{M} = \bigcup_{t \in [a, b]} C(t), \quad \text{with } g_0 \leq a < b \leq 1,$$

$$\text{and } C(t_1) \cap C(t_2) = \emptyset \quad \text{if } t_1 \neq t_2,$$

Geometric modeling: Whitney embedding-Dimension Reduction

Theorem:

Let m be a compact *Hausdorff* n -dimensional manifold, $2 \leq r \leq \infty$, then there is a embedding of m in \mathbb{R}^{2n+1}

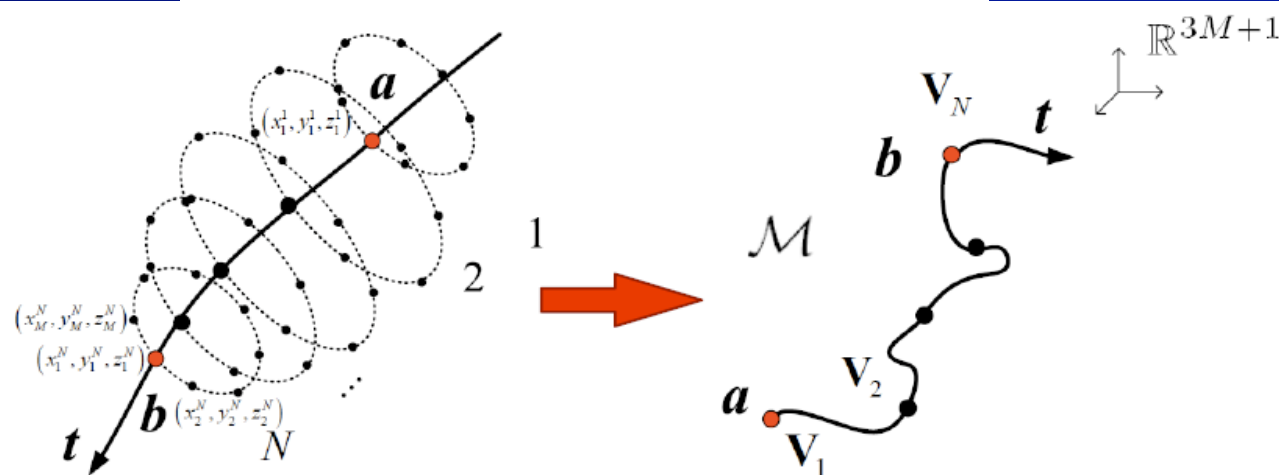
(Constructive proof due to Broomhead and Kirby, SIAM DS 2000)

■ In our problem, we are to embed a in \mathbb{R}^3

[*Aouada, et al. ICIP 08, IEEE Trans. On IP 09*]

Curve Modeling

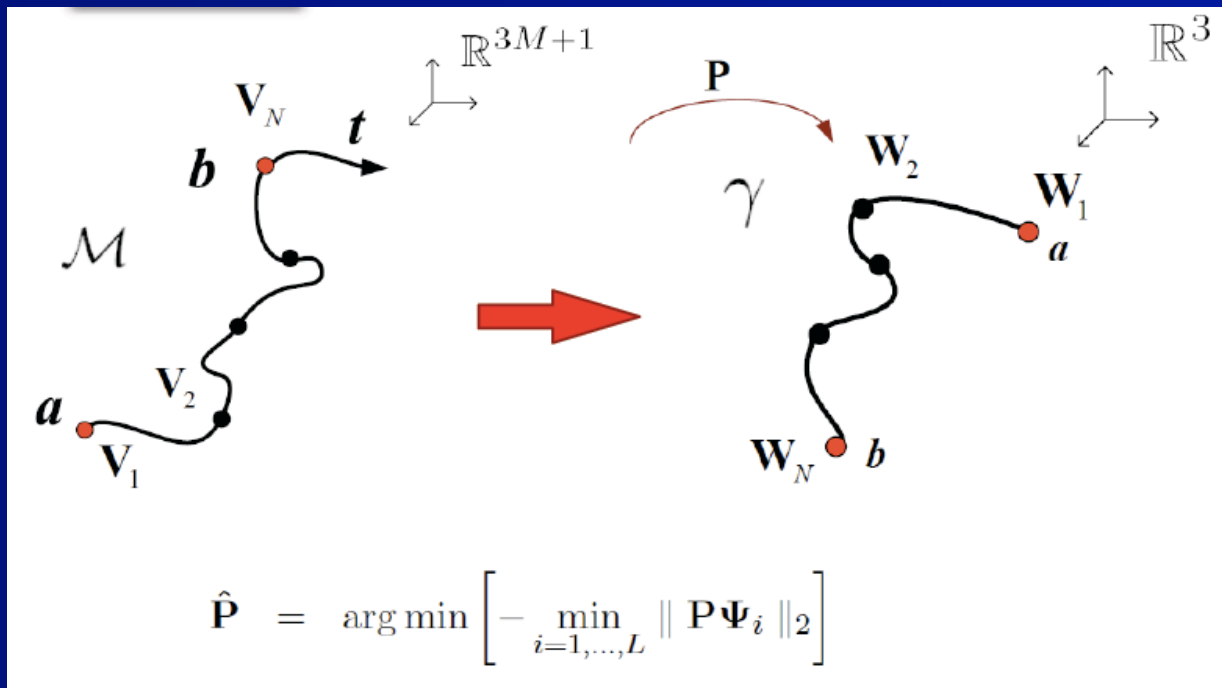
$$\mathbf{V}_i = \left[[\mathbf{V}_1^i]^T \dots [\mathbf{V}_M^i]^T \right]^T \text{ th } \mathbf{V}_j^i = [x_j^i, y_j^i, z_j^i]^T$$



$$\begin{aligned} \Psi &= \left\{ \frac{\mathbf{V}_i - \mathbf{V}_j}{\|\mathbf{V}_i - \mathbf{V}_j\|}, (i, j) \in \{1, \dots, N\}^2 \text{ and } i \neq j \right\} \\ &= \{\Psi_i, \quad i = 1, \dots, L\}, \end{aligned}$$

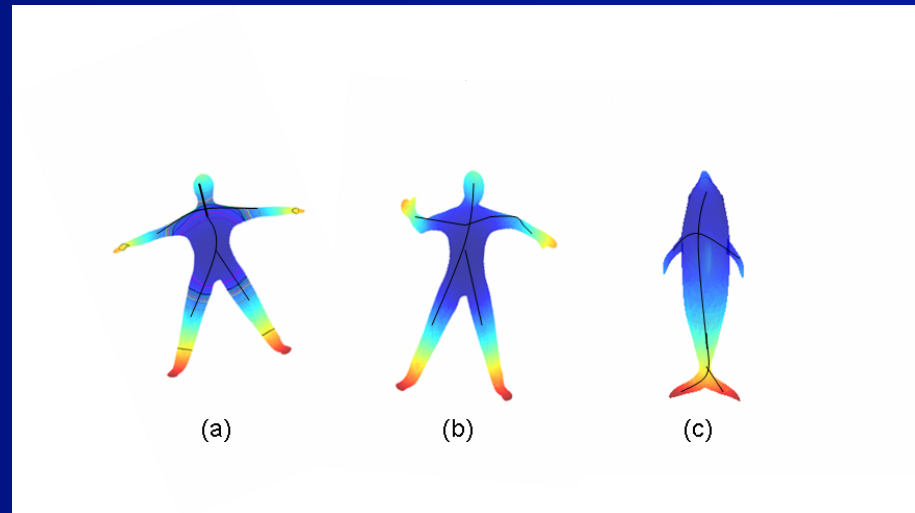
Whitney Embedding

$$\mathbf{W} = \mathbf{P}^T \mathbf{V},$$

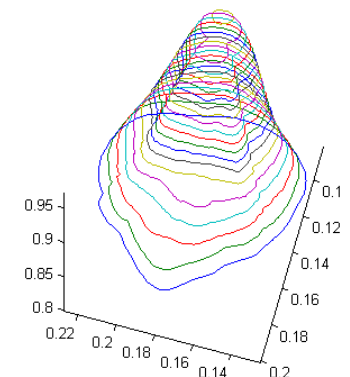
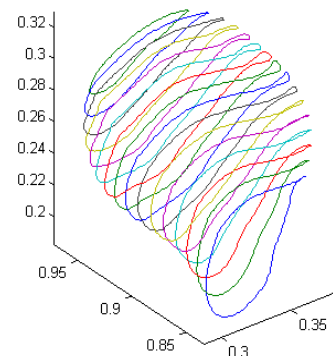
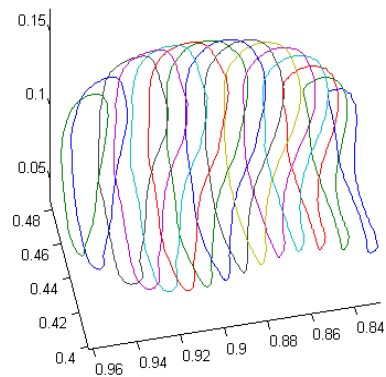
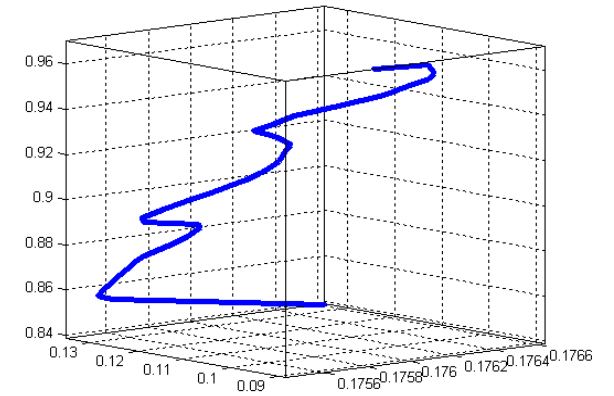
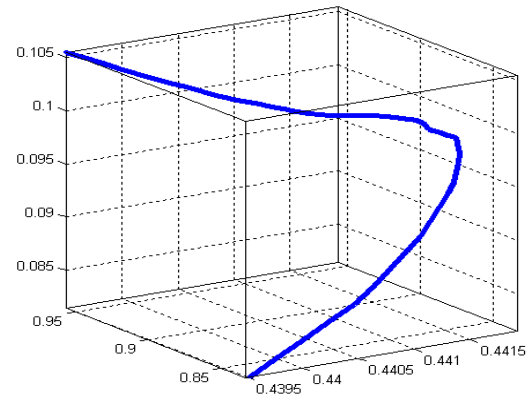
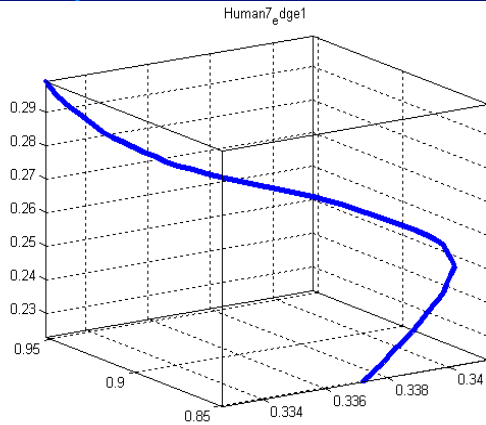


Example illustrating the problem

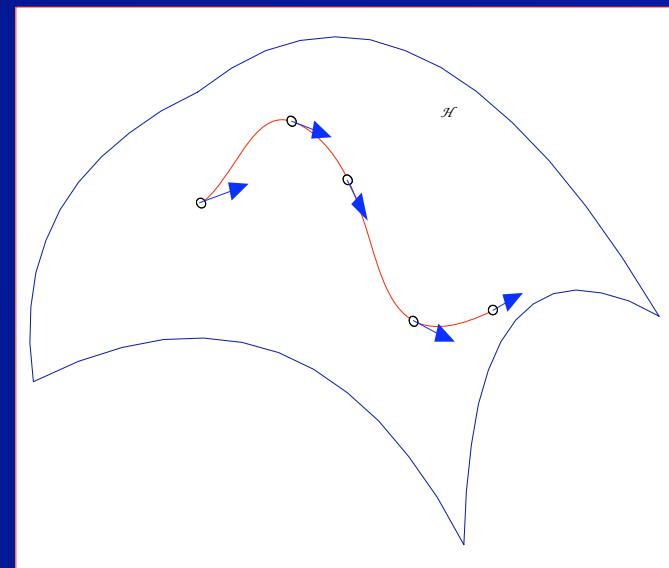
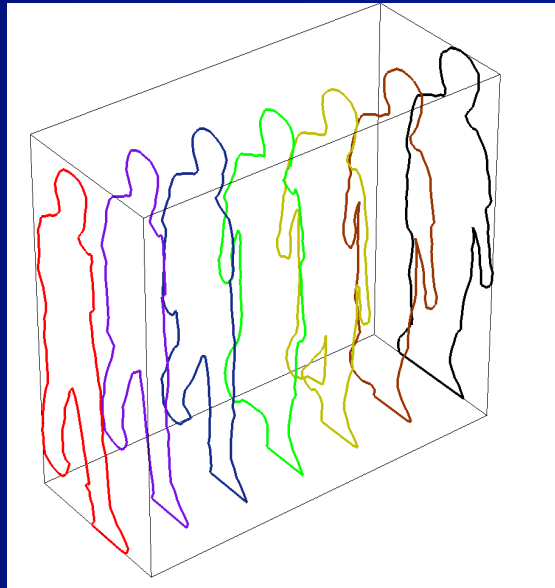
- Distinguishing three objects with the same graphical representation.
- Find 3D curves to assign to each edge.



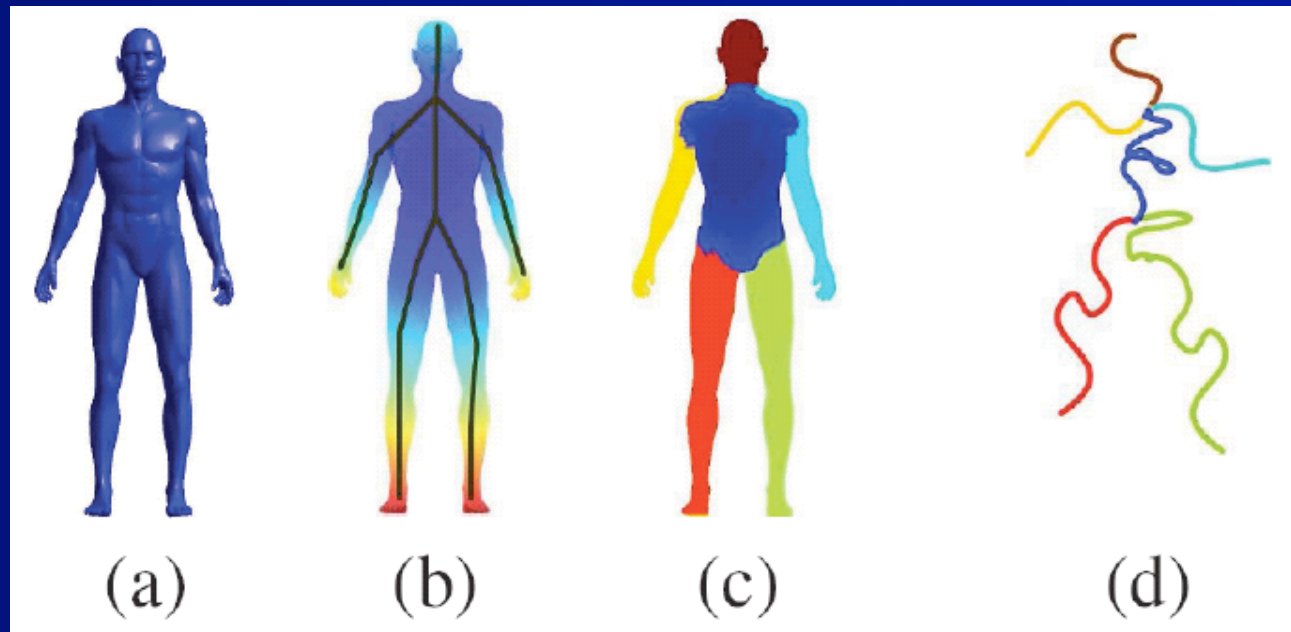
“Modeling of heads”



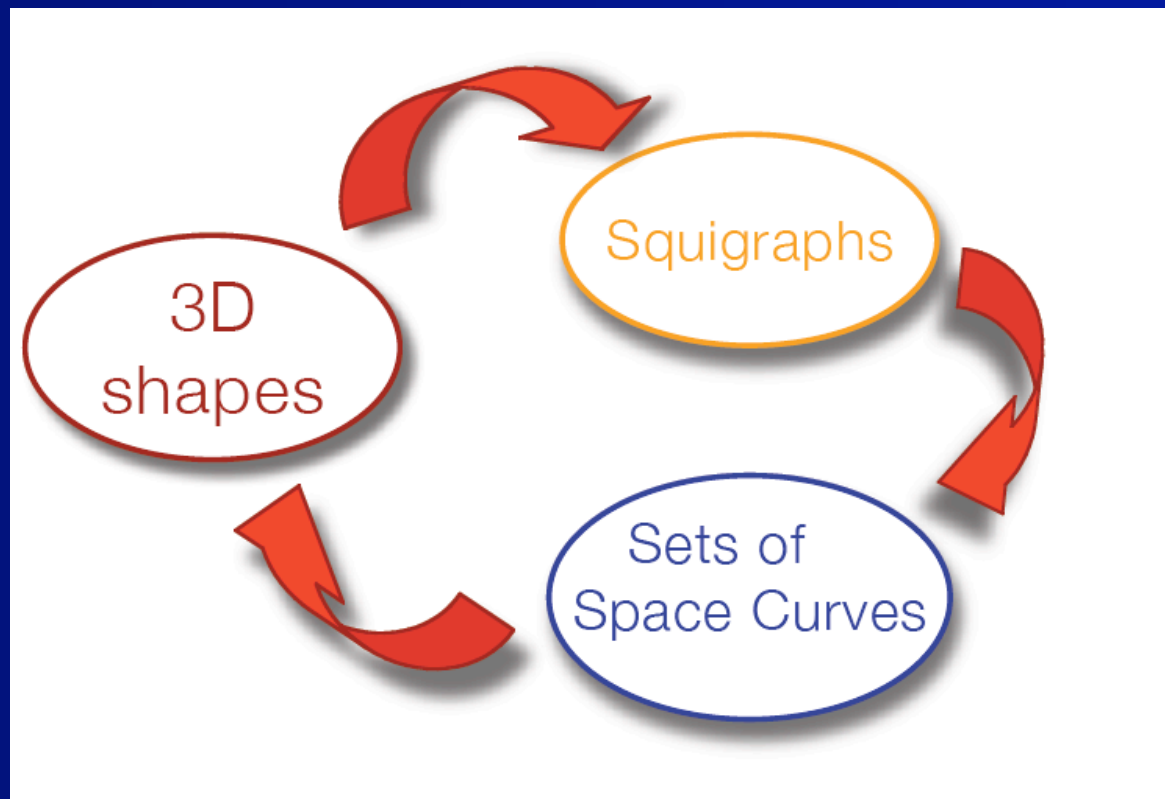
Other Applications



Squigraph Model



Representation



Object comparison/ classification

Registration

- Complicated
- Time consuming



Invariant

- Easy to compute
- Inexpensive to store
- From curve comparison to invariant comparison

Invariants offer a simple alternative !

Geometric Model Matching

- Curves are fundamental blocks in vision and imaging
- Curves undergo geometrical transformations (e.g. biological organs)
- Matching tantamount to coping with variability of characteristic curves

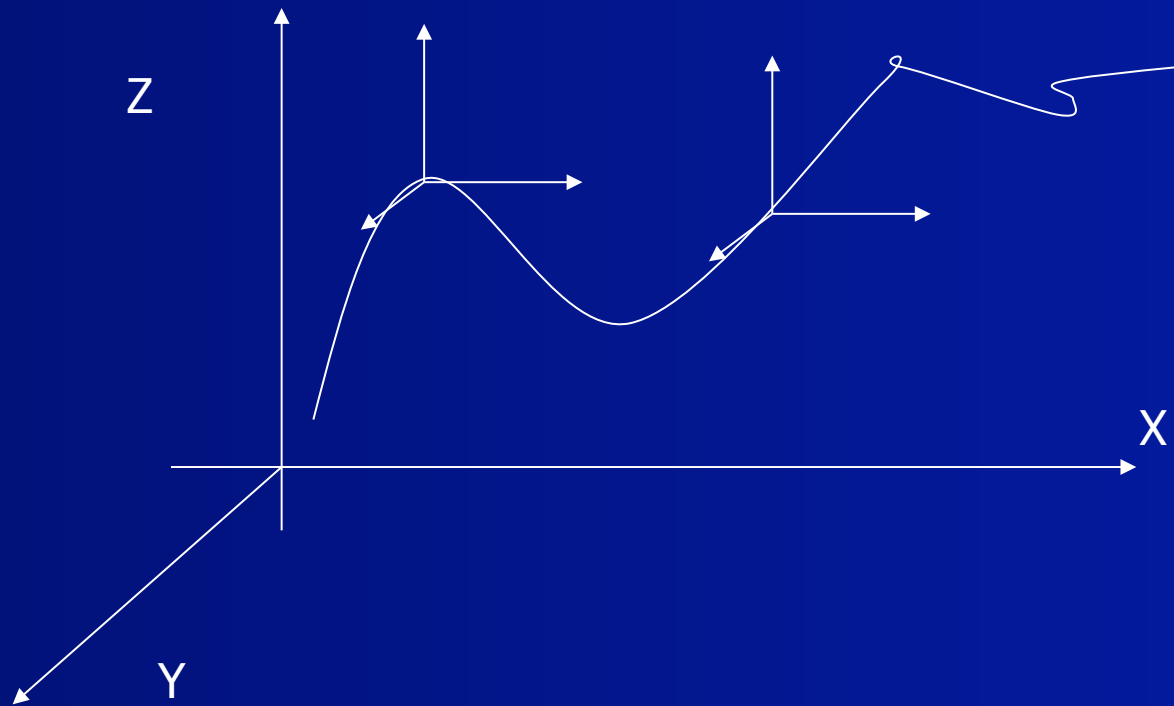


- Invariants (Boutin, Fels-Olver, Kogan, Hickman *et al.*, Manay *et al.*)

Invariants

- Integral Invariants
 - Robust to noise
 - Too costly to make independent of parameterization
 - Hard to compare
- Differential Invariants
 - Easy to compute and to apply
 - Sensitive to noise

Intuitively...invariant



Integral Jet Space

- Special affine action on spatial curves can be prolonged to the action on the integral variables of order $l=i+j$ (with *S. Feng, I. Kogan*)

$$X_{ijk} = \int_{t_0}^t x^i y^j z^k dx, \quad j+k \neq 0 \quad Y_{ijk} = \int_{t_0}^t x^i y^j z^k dy \quad i+k \neq 0$$

$$Z_{ijk} = \int_{t_0}^t x^i y^j z^k dz, \quad i+j \neq 0$$

- The jet space is defined as:

$$(X, Y, Z, Z_{100}, Z_{010}, Y_{100}, Z_{011}, Z_{020}, Z_{101}, Z_{110}, Y_{101}, X_{110}, X_{101}, X_{020})$$

3D Integral Invariants

- We achieve two special affine integral invariants:

$$I_1 = n_1 X + n_2 Z - n_3 Y$$

$$I_2 = 2n_1(XYZ^2 - 3Z_{011}X + 3YZ_{101} - ZZ_{110} - 2ZY_{101}) + n_2(2XYZ - 3XZ_{020} - 6ZX_{020} - 4YZ_{110} - 2YY_{101}) - 2n_3(3YX_{101} - 3ZX_{110} + XZ_{110} - XY_{101})$$

$$n_1 = YZ - 2Z_{010} \quad n_2 = XY - 2Y_{100} \quad n_3 = XZ - 2Z_{100}$$

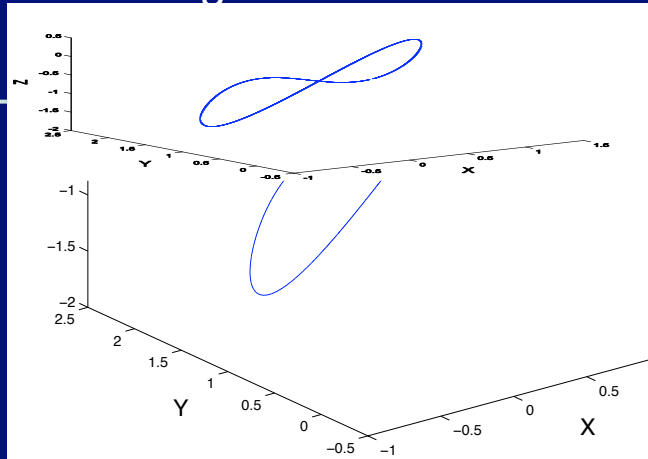
where

- The full affine integral invariant is easily derived as [Feng et al., 09]

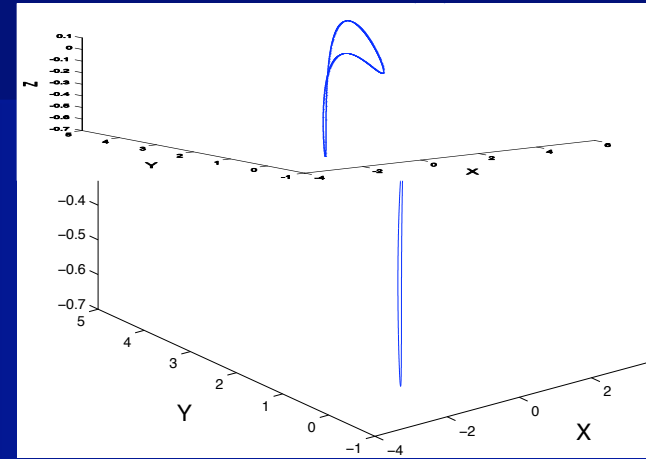
$$I = \frac{I_2}{I_1^2}$$

Examples of invariants

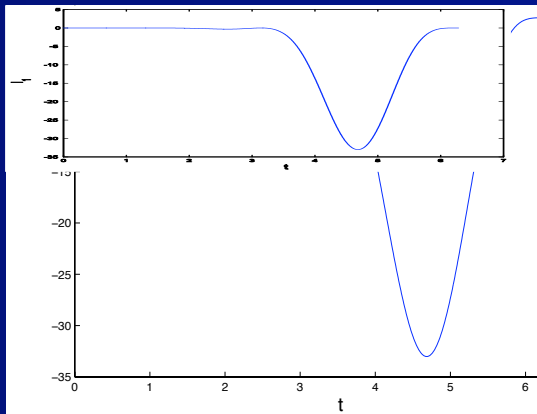
Original Curve



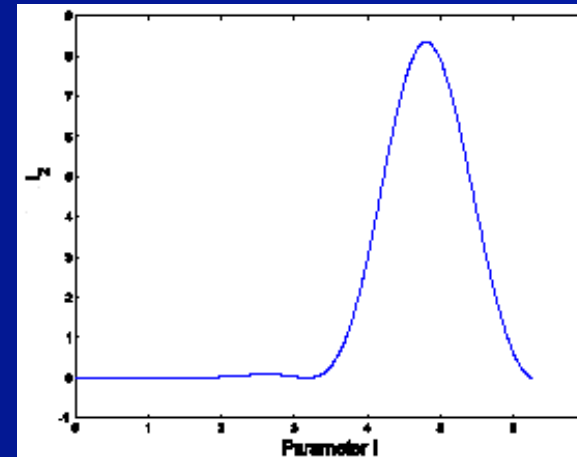
Transformed Curve



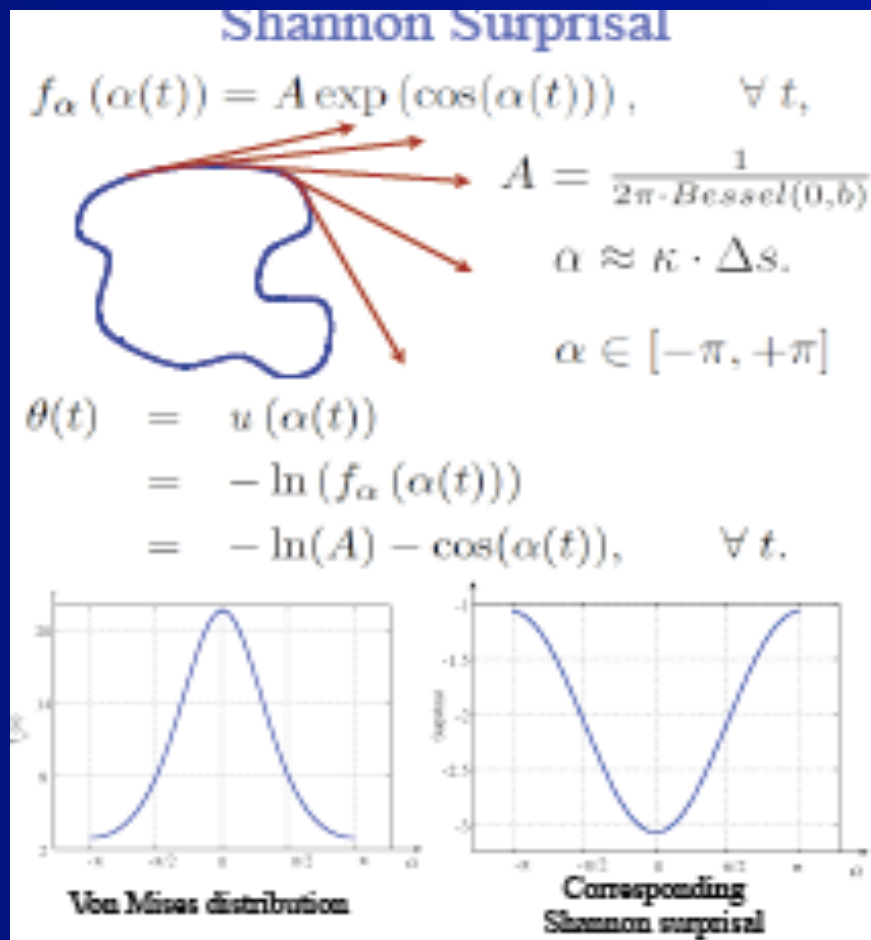
Invariant I1 for both of the curves above



Invariant I2 for both of the curves above



Invariants: Shannon Surprisal



Space Curves

Using Frenet frame to generalize:

$$\begin{aligned}\frac{d\mathbf{T}}{dt} &= \kappa\mathbf{N}, \\ \frac{d\mathbf{N}}{dt} &= -\kappa\mathbf{T} + \tau\mathbf{B}, \\ \frac{d\mathbf{B}}{dt} &= -\tau\mathbf{N},\end{aligned}$$

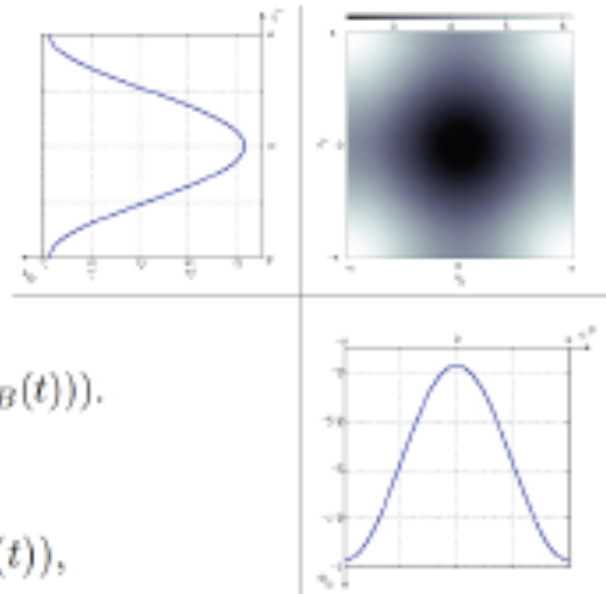


$$\alpha_T(t) \approx \kappa(t) \cdot \Delta t \quad \text{and} \quad \alpha_B(t) \approx \tau(t) \cdot \Delta t.$$

$$f_\alpha(\alpha_T(t), \alpha_B(t)) = A^2 \exp(\cos(\alpha_T(t)) + \cos(\alpha_B(t))).$$

$$\begin{aligned}\theta(t) &= -\ln(f_\alpha(\alpha_T, \alpha_B)) \\ &= -2 \ln(A) - \cos(\alpha_T(t)) - \cos(\alpha_B(t)),\end{aligned}$$

From planar to space curves



Intrinsic Riemannian Metric

- For two given curves γ_1, γ_2 , define an oriented curve $\lambda_\Delta = \lambda_1 - \lambda_2$
- First take form F ,

$$\vec{F} : ([-\pi, \pi])^2 \rightarrow \mathbb{R}^2$$

$$(\alpha_T, \alpha_B) \rightarrow -\ln(f_\alpha(\alpha_T)) \vec{i} - \ln(f_\alpha(\alpha_B)) \vec{j}$$

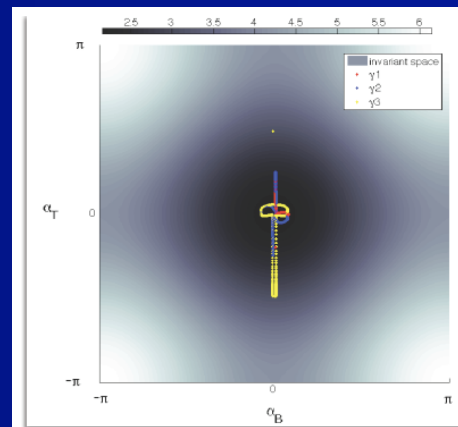
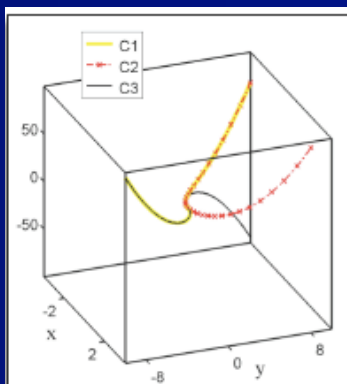
One then defines a projection on $[2\pi]^2$

$$\lambda_\Delta^*(\phi) = \int_{\lambda_\Delta} \phi(F(t)) dt$$

Riemannian Metric

- Define a Flat norm (Vaillant, Glaunes 07)

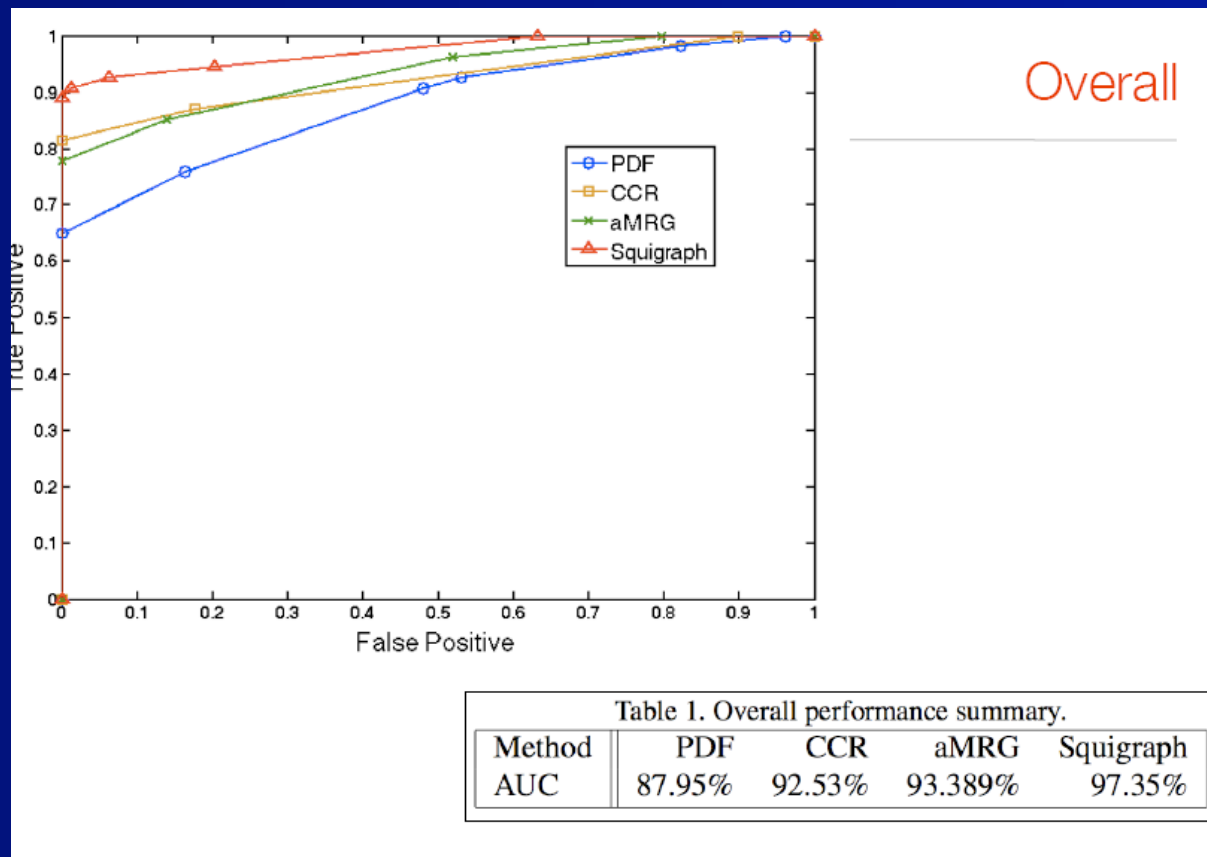
$$D(\gamma_1, \gamma_2) = \mathcal{F}(\lambda_1^* - \lambda_2^*) \triangleq \sup \{ \lambda_\Delta^*(\phi) : \|d\phi\| \leq 1, \forall \|\phi\| \leq 1 \}$$



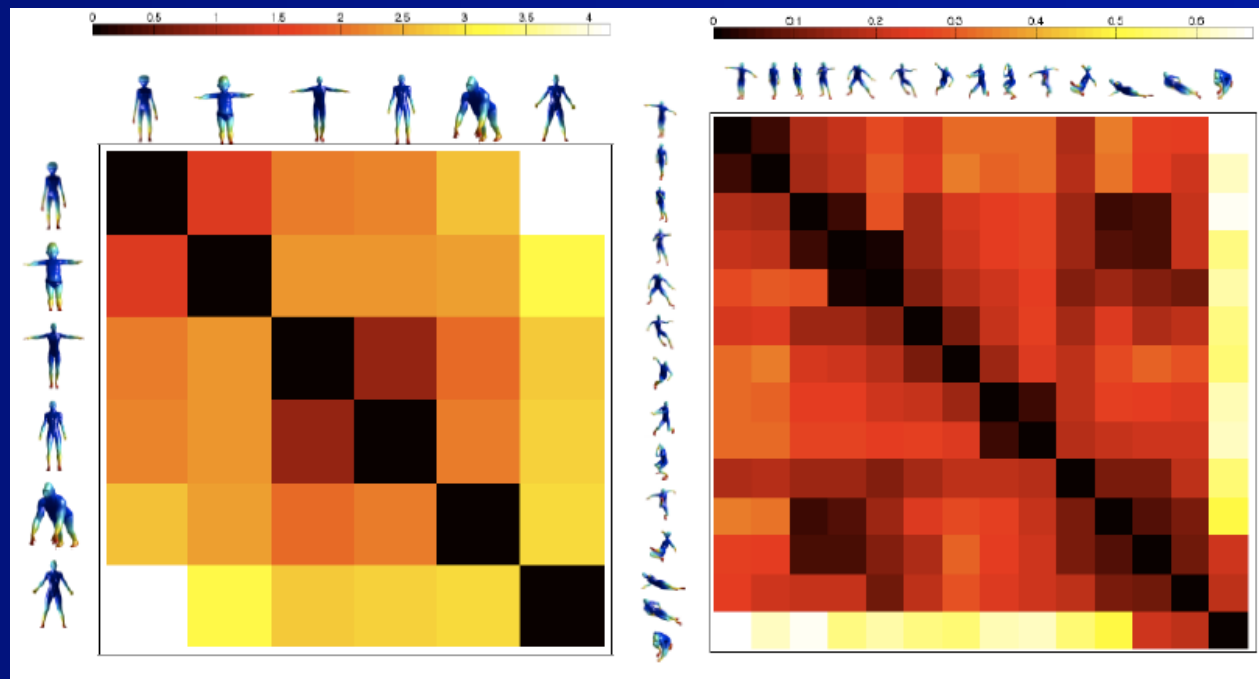
Comparison of complex shapes



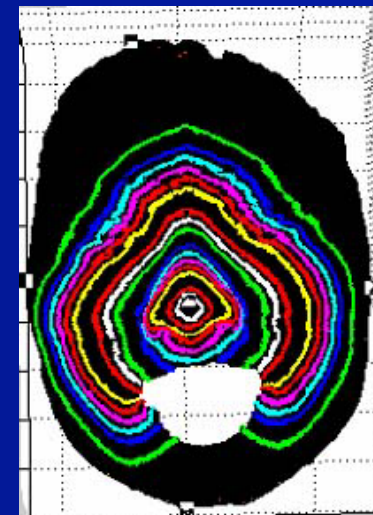
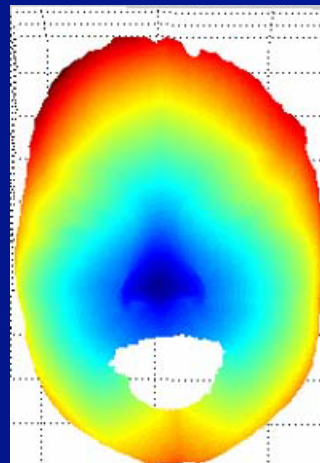
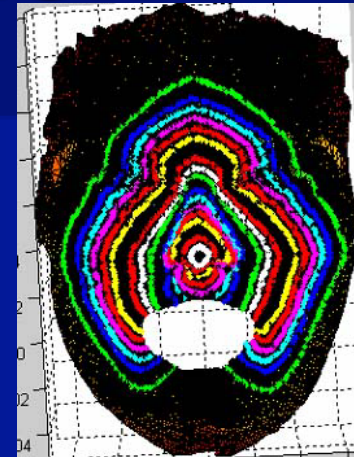
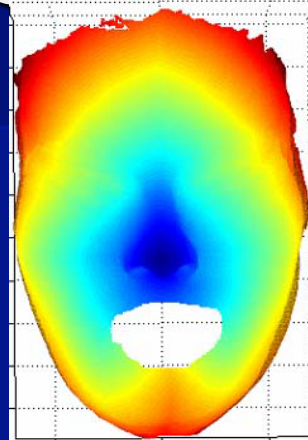
Performance Evaluation



Bipedal Shapes



Face Representation



Conclusion/Ongoing Work

- Framework for 3D shape modeling
- Other theoretical issues
 - Robustness issues
 - Sampling in 3D
- Other application avenues
- Experiment with real data