

# Shape Correspondence on the Unit Sphere

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Preparing for Submission to Symposium on Geometry Processing 2019

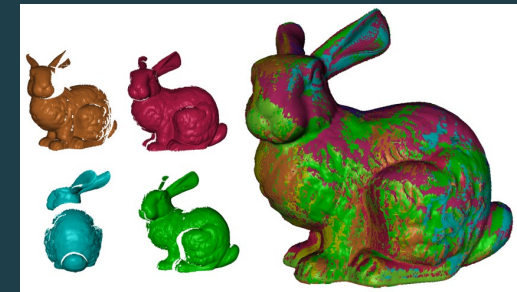
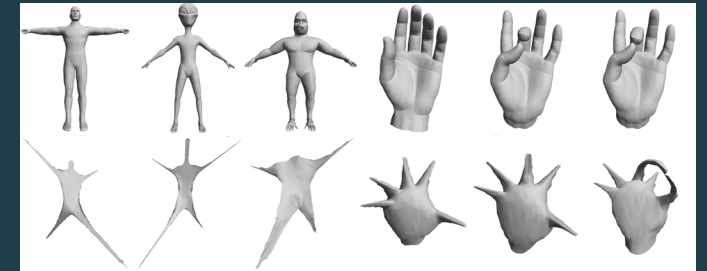
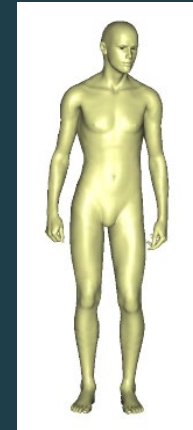
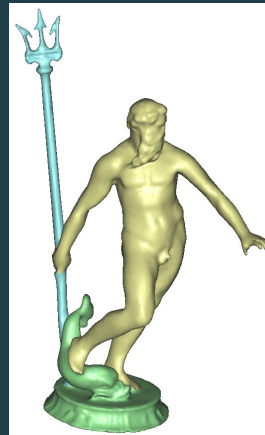
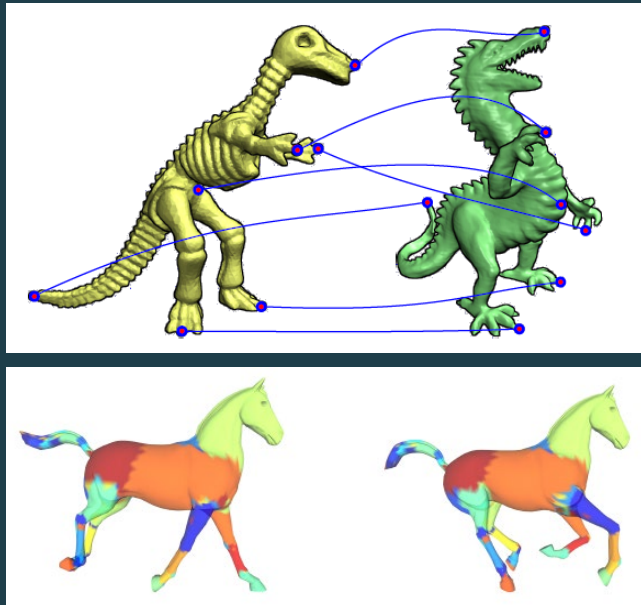
<sup>1</sup>Johns Hopkins University



- Problem Description
- Related Works
- Review of Mathematical Tools
- Problem Formulation
- Our Approach
- Current Results and Demo
- Q&A and Feedback

- Shape Correspondence

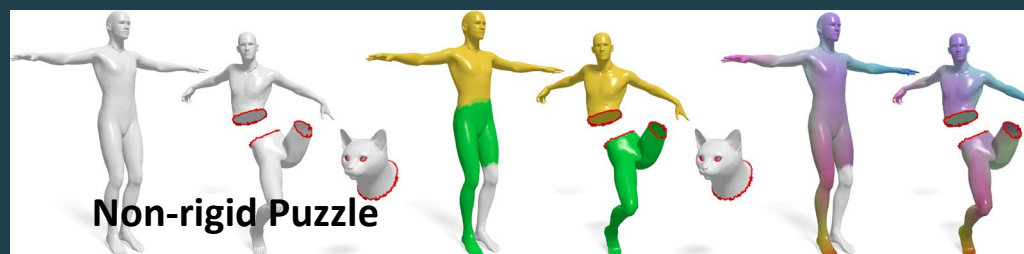
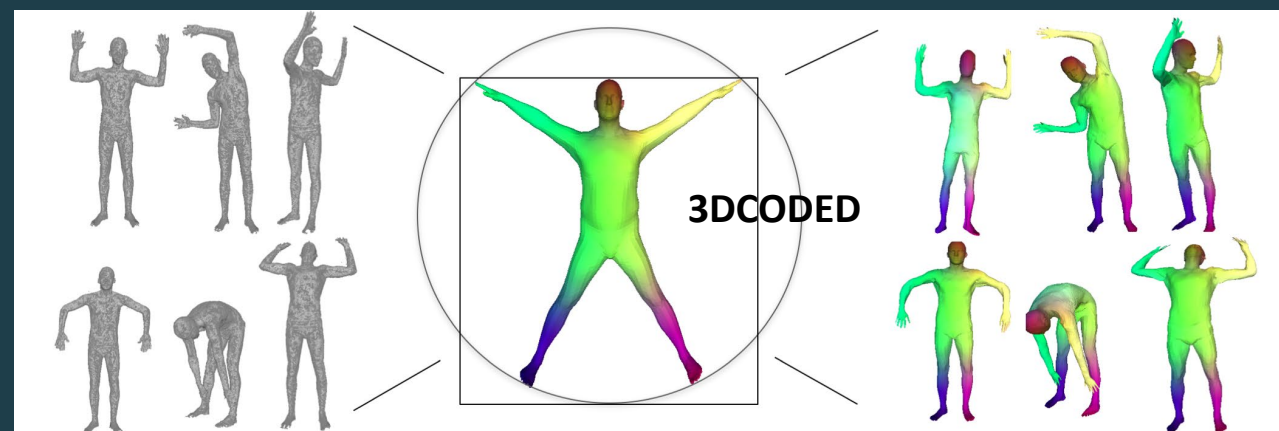
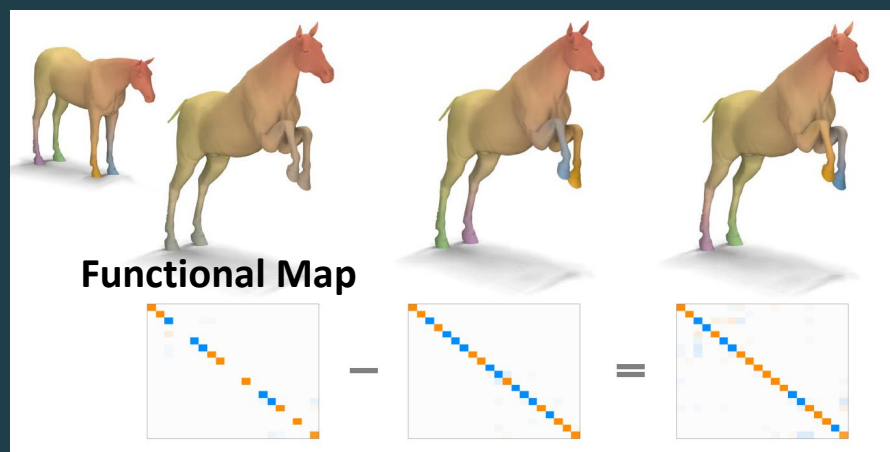
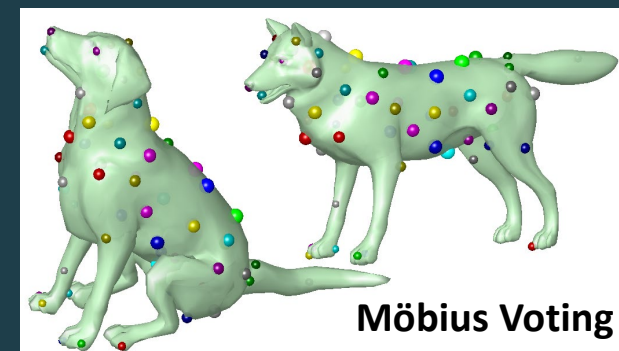
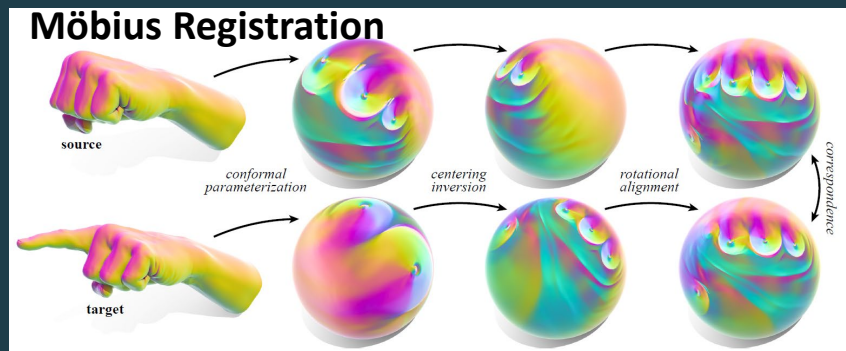
- Given two shapes  $S_1, S_2 \subset \mathbb{R}^3$ , we want to find meaningful map  $\Xi : S_1 \rightarrow S_2$  such that  $\{(s, \Xi(s)) | s \in S_1, \Xi(s) \in S_2\}$  give us a meaningful correspondence between them



\*Pictures borrowed from: O. Kaick, D. Cohen-Or et al., "A Survey on Shape Correspondence", in *CGF 2011*

- Output:
  - **Full** vs. Partial
  - **Dense** vs. Sparse
- Objective function:
  - Rigid vs. **Deformation**
- Approach:
  - **Fully-automatic** vs. Semi-automatic
  - Global vs. **Local**
  - **Pairwise** vs. Groupwise

- Möbius Voting
- Möbius Registration
- Functional Map
- Non-rigid Puzzle
- 3DCODED

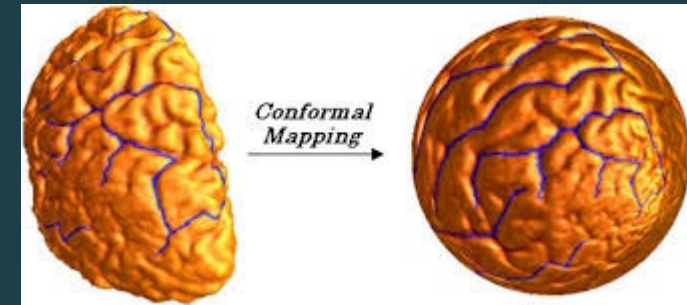
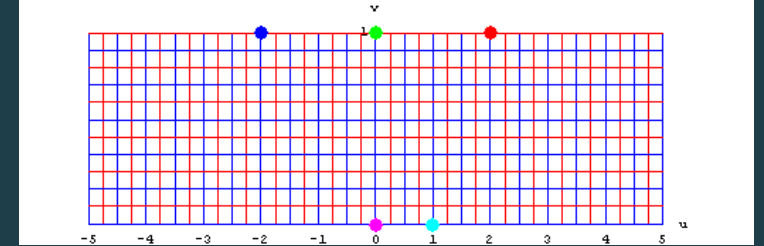
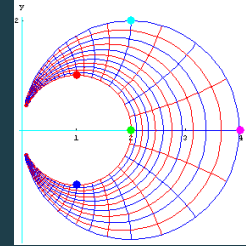


\*Pictures borrowed from the captioned publications



# Möbius Transformation

- Aka. Homographies
- General form in  $\mathbb{C}$  ( $ad - bc \neq 0$ ):
  - $f(z) = \frac{az+b}{cz+d}$
- Matrix form  $\mathbb{C}^{2 \times 2}$  ( $\det \neq 0$ ):
  - $\begin{bmatrix} a & b \\ c & d \end{bmatrix}$
- It's conformal (angle-preserved): map line to a line or circle, and map circle to a line or circle
- Area ratio before/after transformation: conformal factors
- A tool to map genus zero shape ( $S$ ) to the unit sphere ( $\mathbb{S}^2$ )
  - Denote by  $\Phi: S \rightarrow \mathbb{S}^2$

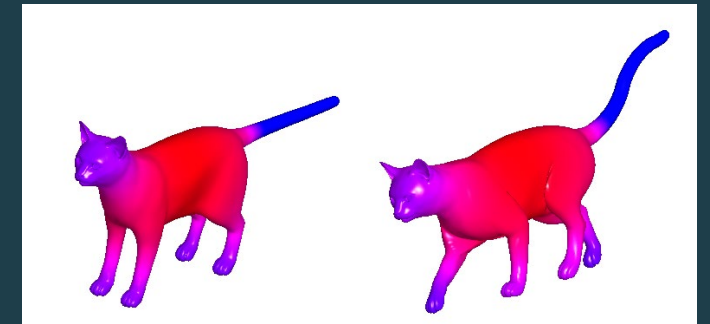
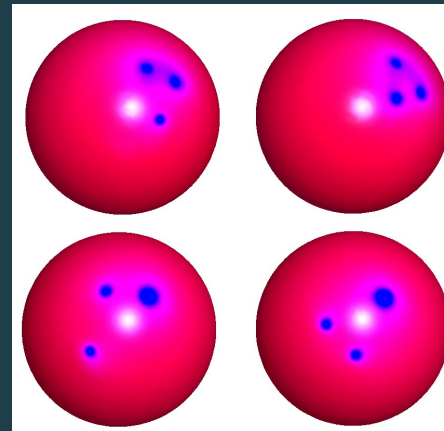
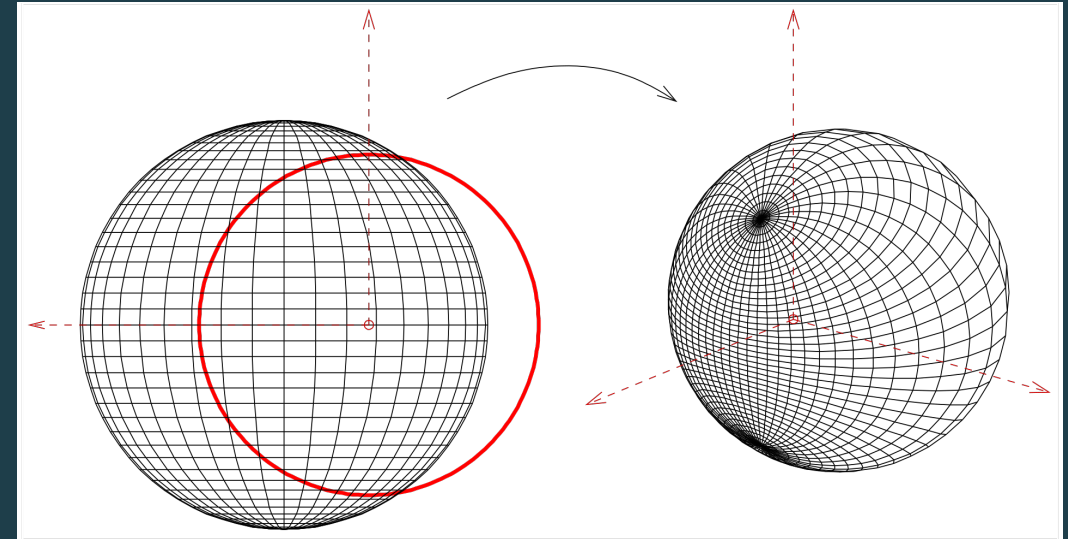


\*Pictures borrowed from "Landmark constrained genus zero surface conformal mapping and its application to brain mapping research, IMACS 2006"



# Möbius Registration via Spherical Inversion

- $\eta_c: \mathbb{S}^2 \rightarrow \mathbb{S}^2$ 
  - $\eta_c := (1 - |c|^2) \frac{p+c}{|p+c|^2} + c$
- Unique Möbius transformation (up to Rotation) by finding  $\eta_c$  such that the average of the conformal factors is zero
- i.e. center of mass is at the origin
- Möbius registration = aligning conformal factors for rotation







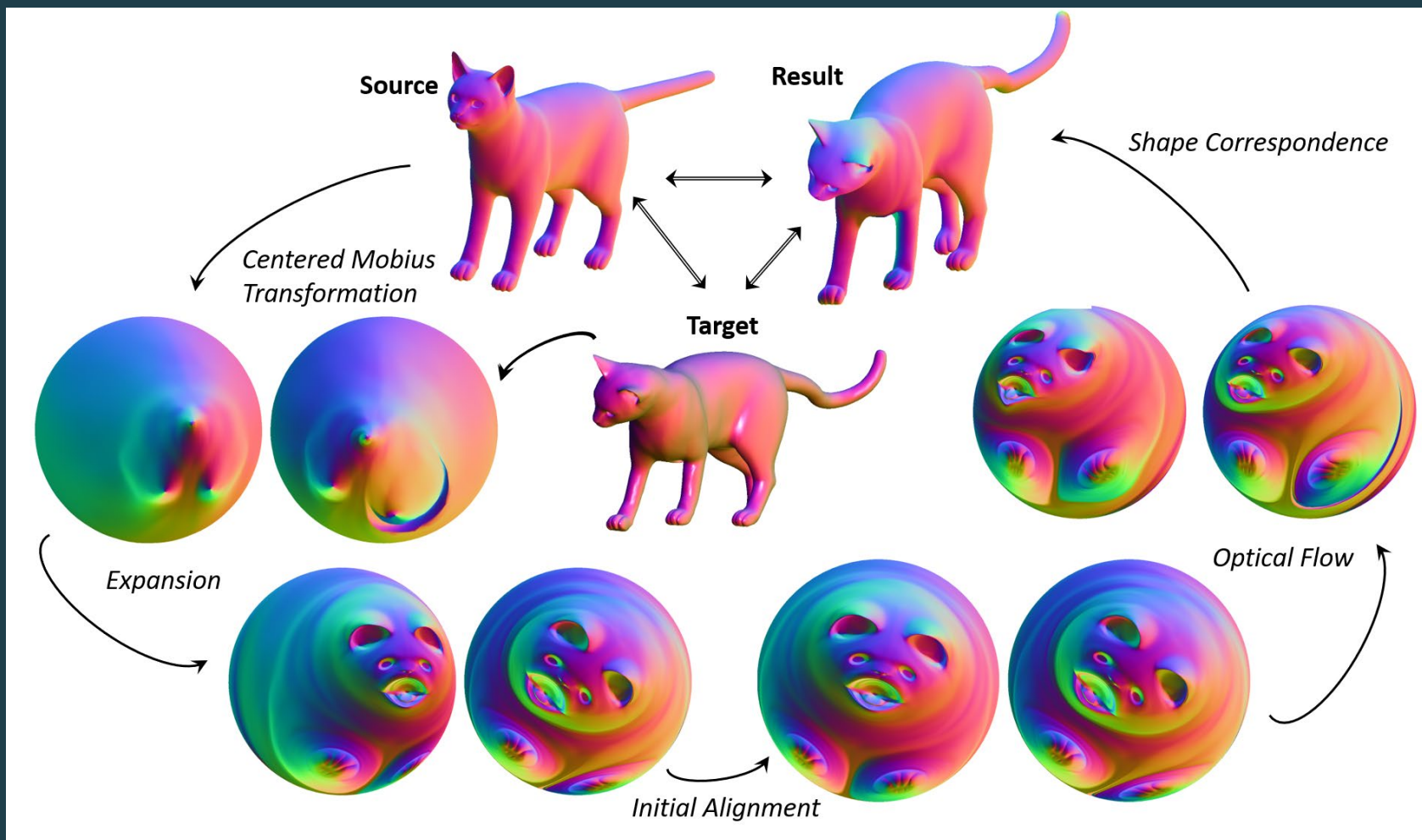
- A generalized version of optical flow on surface is defined as follows: given two signals  $f_i: S_i \subset \mathbb{R}^3 \rightarrow \mathbb{R}$  on two surface meshes  $S_i$  with correspondence, we can estimate an optical flow  $\vec{v}$  by minimizing:

$$E(\vec{v}) = \int_{\mathbb{R}^3} \|\langle \nabla f_1(p), \vec{v}(p) \rangle - \delta(p)\|^2 + \|\langle \nabla f_2(p), \vec{v}(p) \rangle - \delta(p)\|^2 + \epsilon \|\nabla \vec{v}(p)\|^2 dp$$

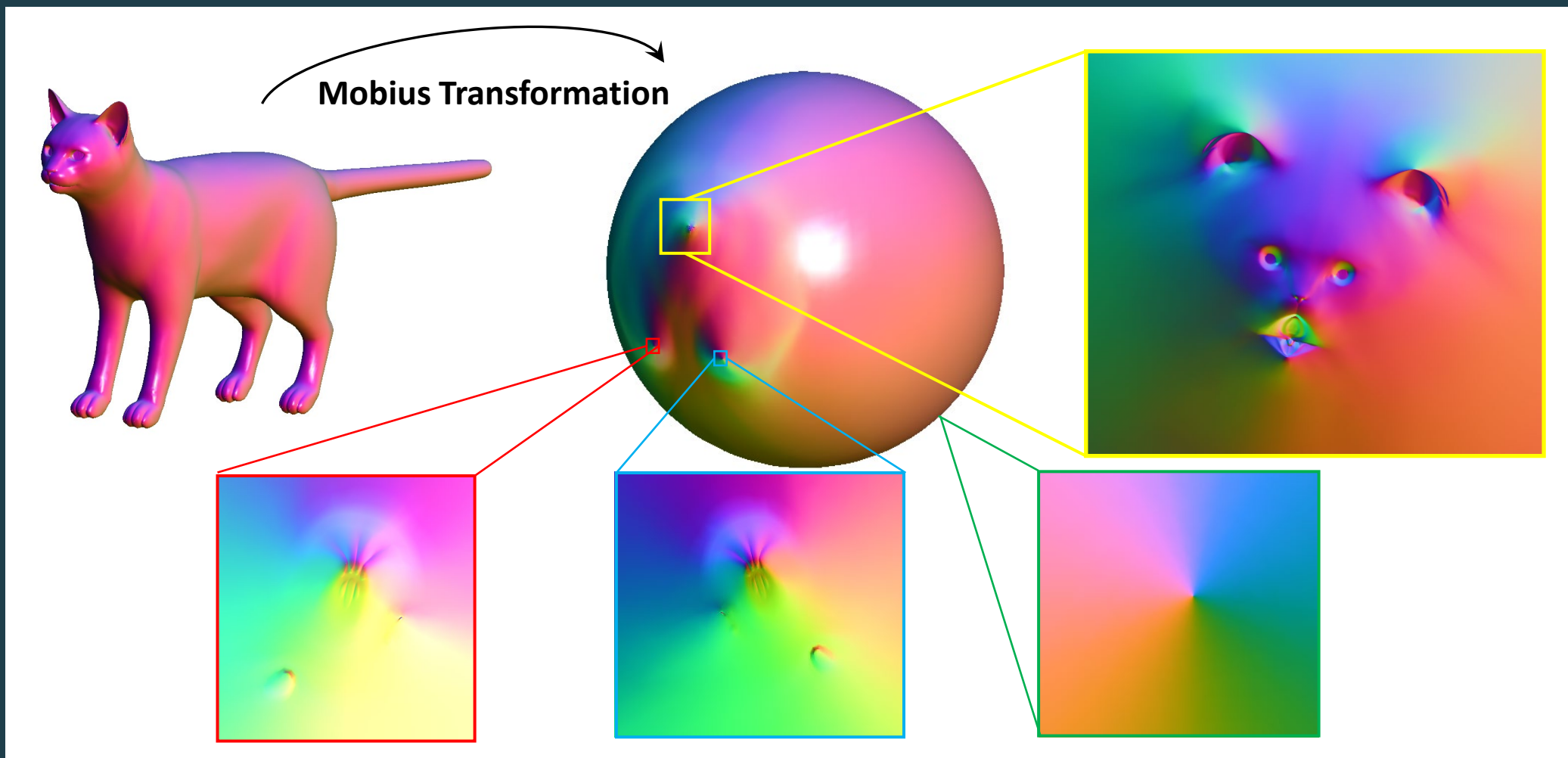
where  $\langle \nabla f_i(p), \vec{v}(p) \rangle$  is the advection of the signal between two meshes,  $\delta(p) = f_1(p) - f_2(p)$  is the difference between two signals, and  $\|\nabla \vec{v}(p)\|$  is the smoothness term to regularize the vector field  $\vec{v}$



- Given two shapes  $S_1, S_2 \subset \mathbb{R}^3$ , corresponding signals  $f_i : S_i \rightarrow \mathbb{R}^d$
  - Compute and sample  $f_i$  of  $S_i$  on  $\mathbb{S}^2$  via unique centered Möbius transformation, namely  $\Phi_i : S_i \rightarrow \mathbb{S}^2$ ,  
i.e. pullback to  $\mathbb{S}^2$  -  $g_i := f_i \circ \Phi_i^{-1} : \mathbb{S}^2 \rightarrow \mathbb{R}^d$
  - Perform initial alignment (a rotation) that minimize
    - $E(R) = \min_R \int_{\mathbb{S}^2} \|g_1 \circ R(p) - g_2(p)\|^2 dp$
  - Estimate optical flow that minimize
    - $E(\vec{v}) = \int_{\mathbb{S}^2} \|\langle \nabla g_1 \circ R(s), \vec{v}(s) \rangle - \delta(s)\|^2 + \|\langle \nabla g_2(s), \vec{v}(s) \rangle - \delta(s)\|^2 + \epsilon \|\nabla \vec{v}(s)\|^2$
- where  $\delta(s) = g_1 \circ R(s) - g_2(s)$



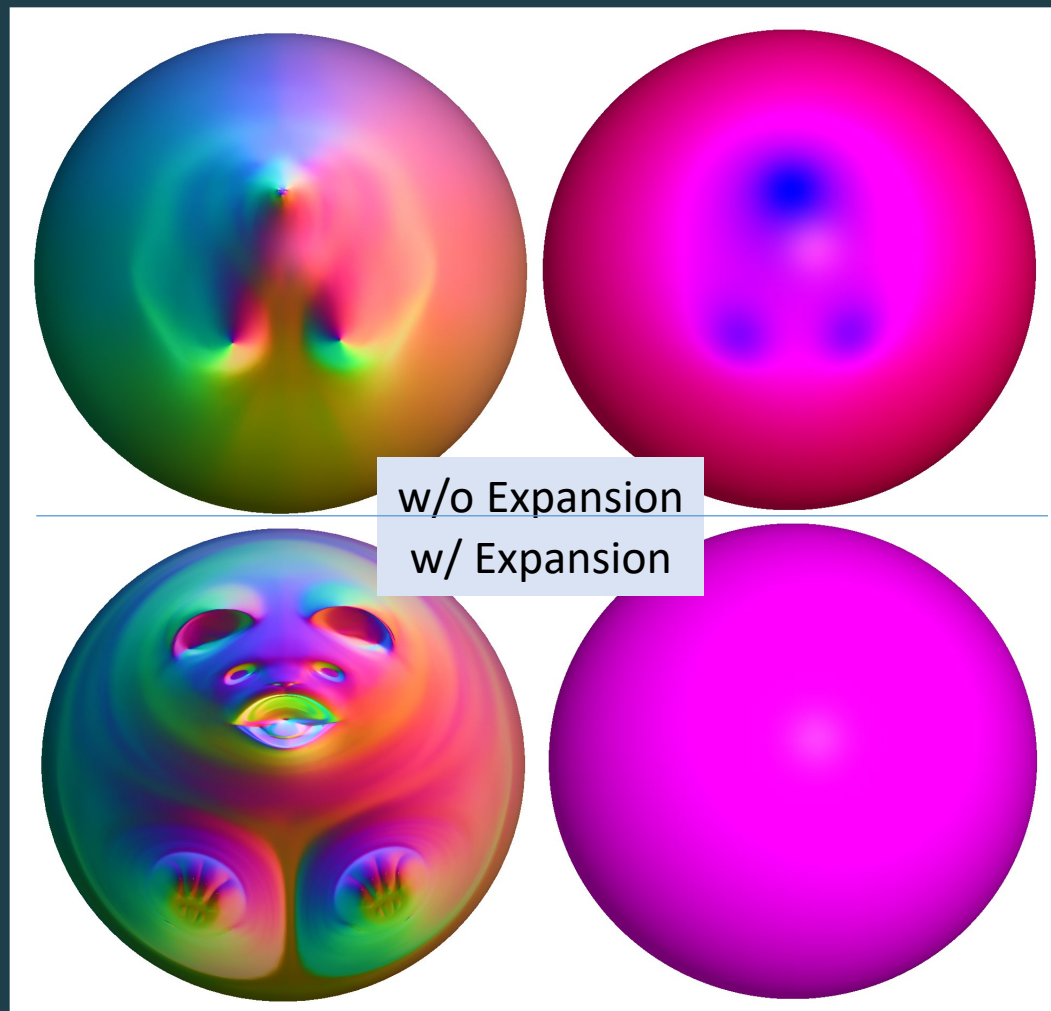
# Step 1 – Möbius Transformation



## Step 2 – Expansion

- Problem: dense region is too packed → hard to discretize
- Idea: expand the dense region out and shrink the sparse region
- How? Using Poisson Equation
- Given a conformal factor  $\phi(x)$ , we would like to uniformized the values
- Goal: find a vector field  $\vec{v}$  such that  $\phi(x + t\vec{v}(x))$  is uniform
- Solve:  $\Delta\phi(x + t\vec{v}(x)) = \frac{d}{dt}\phi(x + t\vec{v}(x))$  at  $t = 0$
- Solution:  $\vec{v}(x) = -\frac{2\Delta\phi}{\phi} = -2\Delta\log\phi$

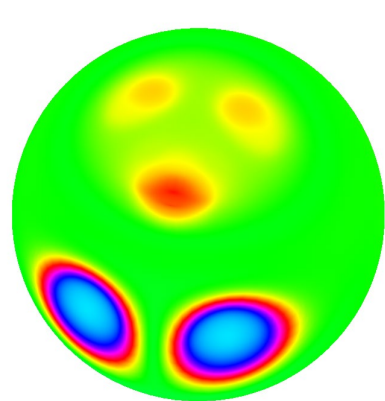
# Expansion Result



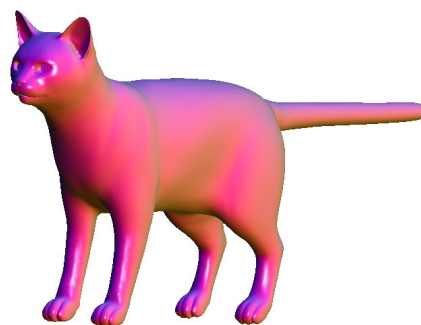
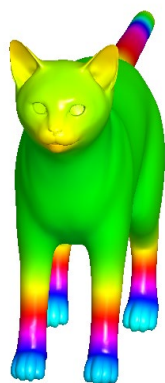
- Isometric invariant signatures
  - Heat Kernel Signature (HKS)
- Eigenvalues and Eigenvectors of Generalized Eigenvalue Decomposition of the Shape Operator
  - Solving  $Sx = \lambda Mx$  where  $S$  and  $M$  are the stiffness and mass matrices
  - Algorithm: Arnoldi iteration
- Let's  $\alpha_i$  be the eigenvector associated with eigenvalue  $\lambda_i$
- HKS is defined by:
  - $h_t(x, y) = \sum_{i=0}^{\infty} e^{-\lambda_i t} \alpha_i(x) \alpha_i(y)$



## Step 3 – Computed Signals



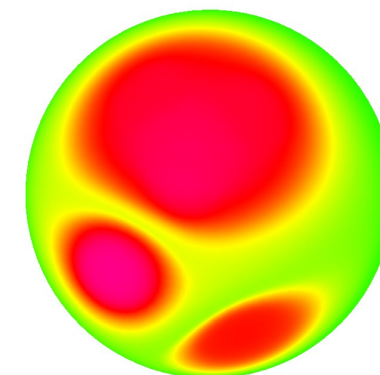
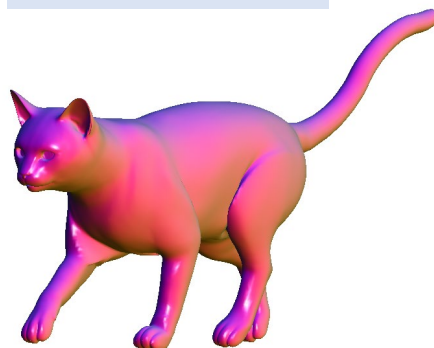
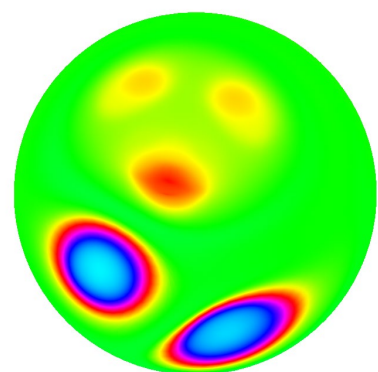
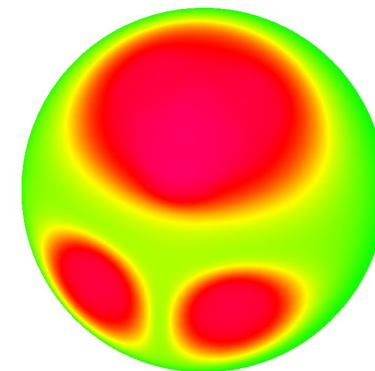
HKS at 0.2121



Input Shapes



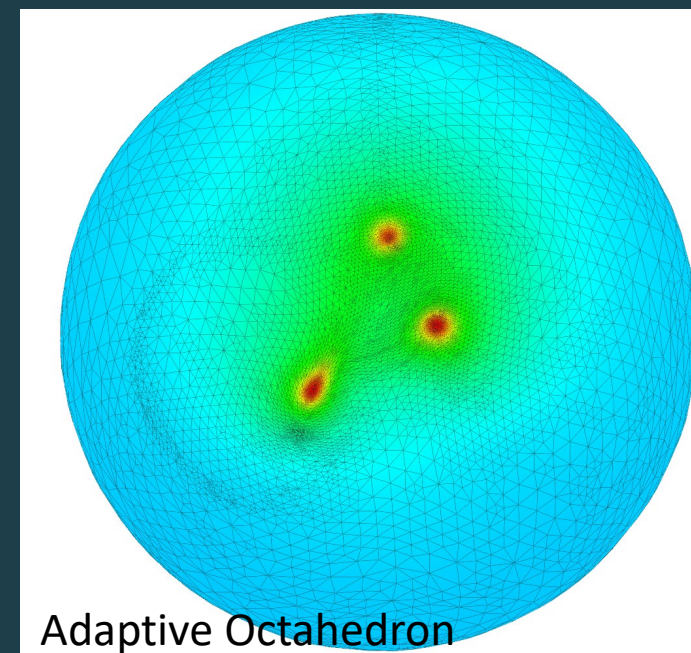
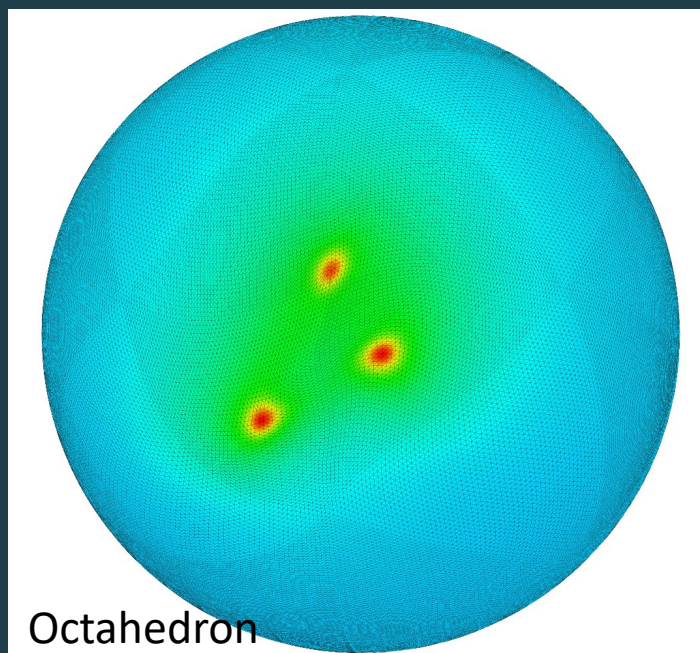
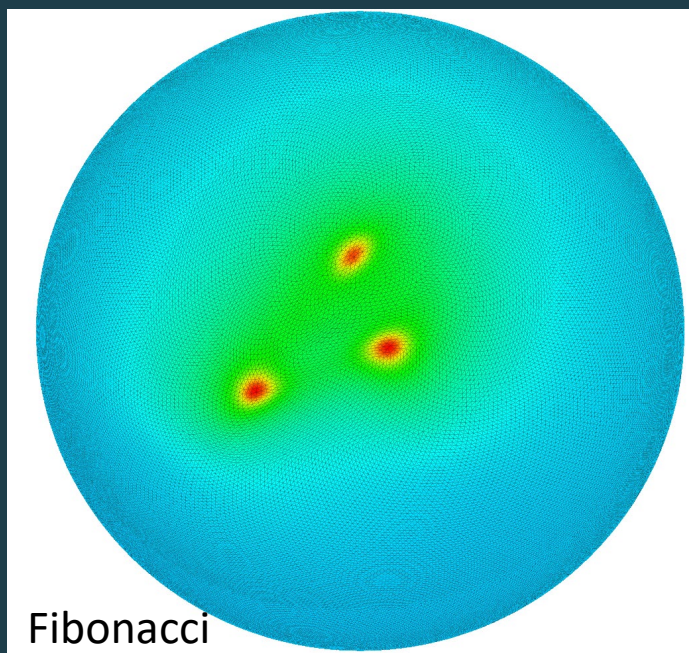
HKS at 0.02





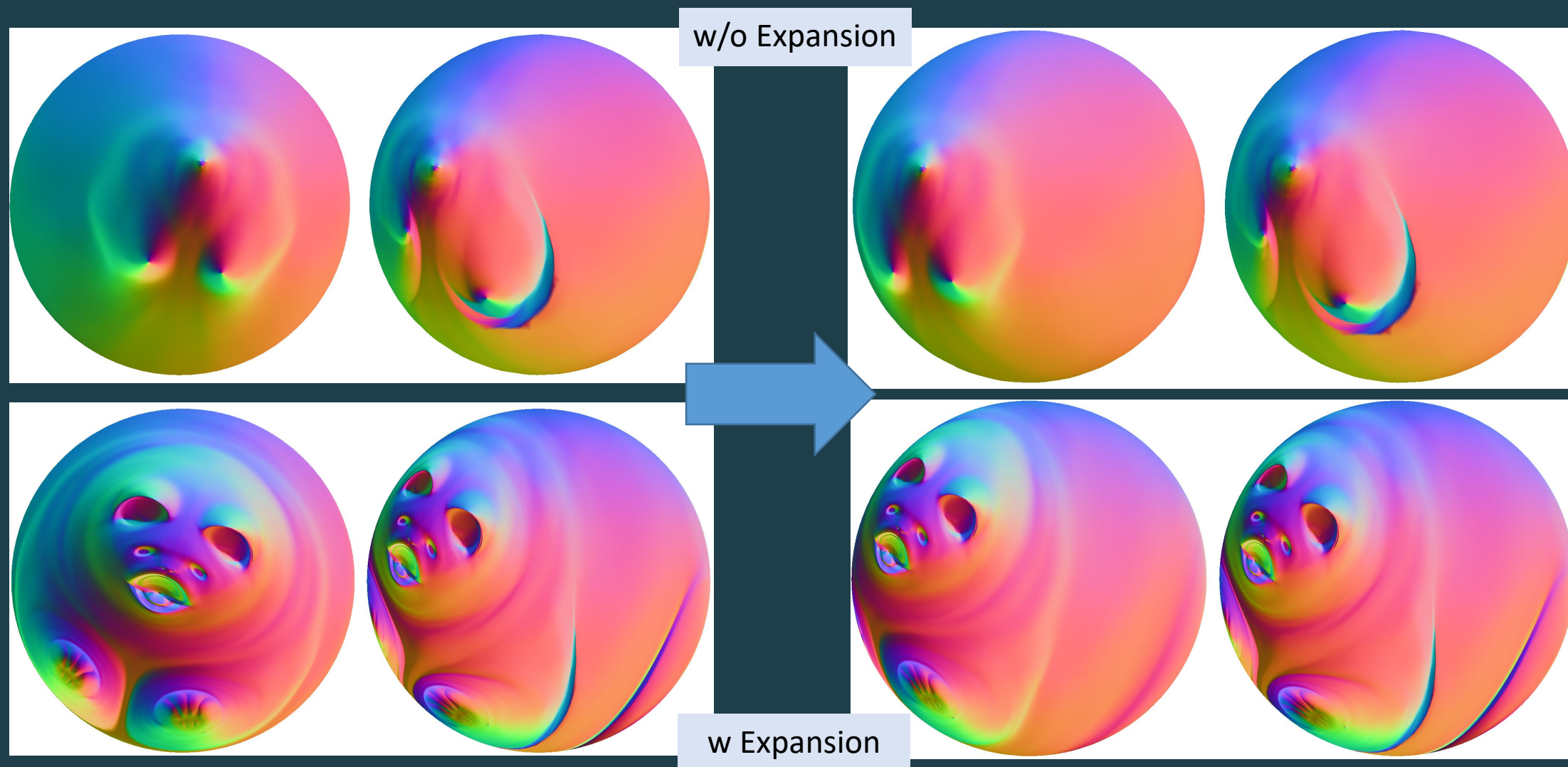
## Step 4 – Sampling on the Sphere

- Fibonacci Sampling
- Octahedron Sampling
- Adaptive Octahedron Sampling



- Harmonic Transform for efficient maximizing of correlation
  - $E(\rho_R) = \min_{\rho_R} \|g_1 - \rho_R(g_2)\|^2$
  - $E(\rho_R) = \min_{\rho_R} \|g_1\|^2 + \|g_2\|^2 - 2\langle g_1, \rho_R(g_2) \rangle$
  - $E(\rho_R) = \max_{\rho_R} \langle g_1, \rho_R(g_2) \rangle$
- It is equivalent to correlation, which can be done efficiently using Harmonic Transform for the sphere

# Alignment Result





# Initial Correspondence

w/o Expansion



w Expansion

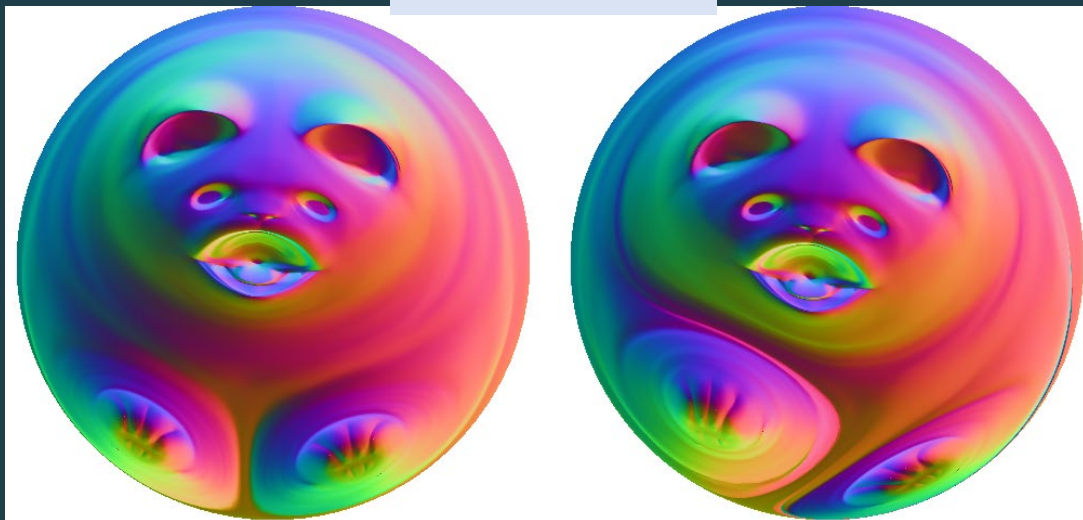


## Step 5 – Surface Optical Flow

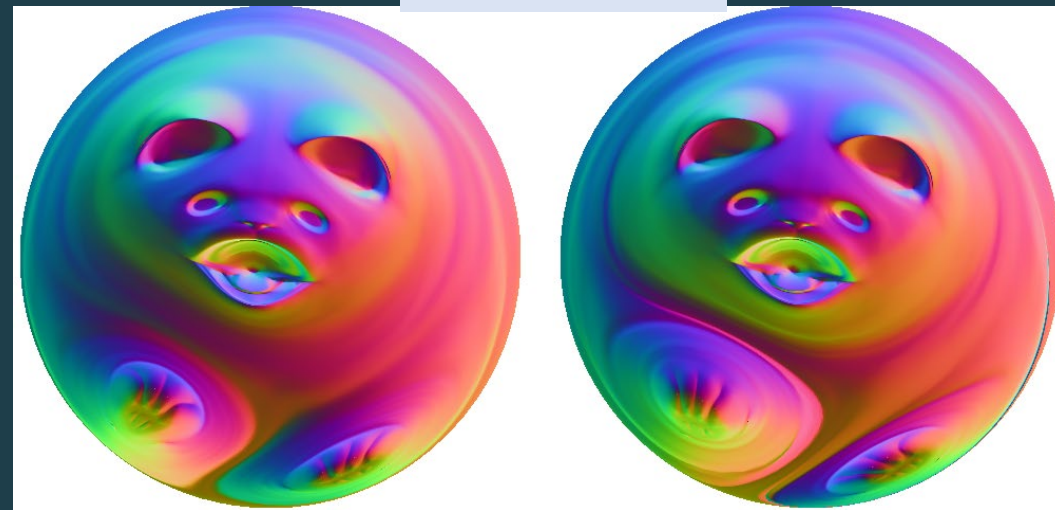
- Recall:
  - $E(\vec{v}) = \int_{\mathbb{S}^2} \|\langle \nabla g_1 \circ R(s), \vec{v}(s) \rangle - \delta(s) \|^2 + \|\langle \nabla g_2(s), \vec{v}(s) \rangle - \delta(s) \|^2 + \epsilon \|\nabla \vec{v}(s)\|^2$
- Signals  $(g_i)$  are HKSs at different time steps
  - Isometric invariant
- But: shapes also undergo non-isometric deformation
- Additional constraint:
  - $E(\vec{v}) = \int_{\mathbb{S}^2} \sum_{i=0,1} \|\langle \nabla g_i(s), \vec{v}(s) \rangle - \delta(s) \|^2 + \epsilon \|\nabla \vec{v}(s)\|^2 + \gamma \|\nabla \cdot \vec{v}(s)\|^2$
  - (With abuse of notation, we write  $\nabla g_1(s) = \nabla g_1 \circ R(s)$ )
  - Meaning: Make the vector field preserve the triangle area (as  $\text{div} = 0$ )

## Step 5 – Surface Optical Flow

Before



After

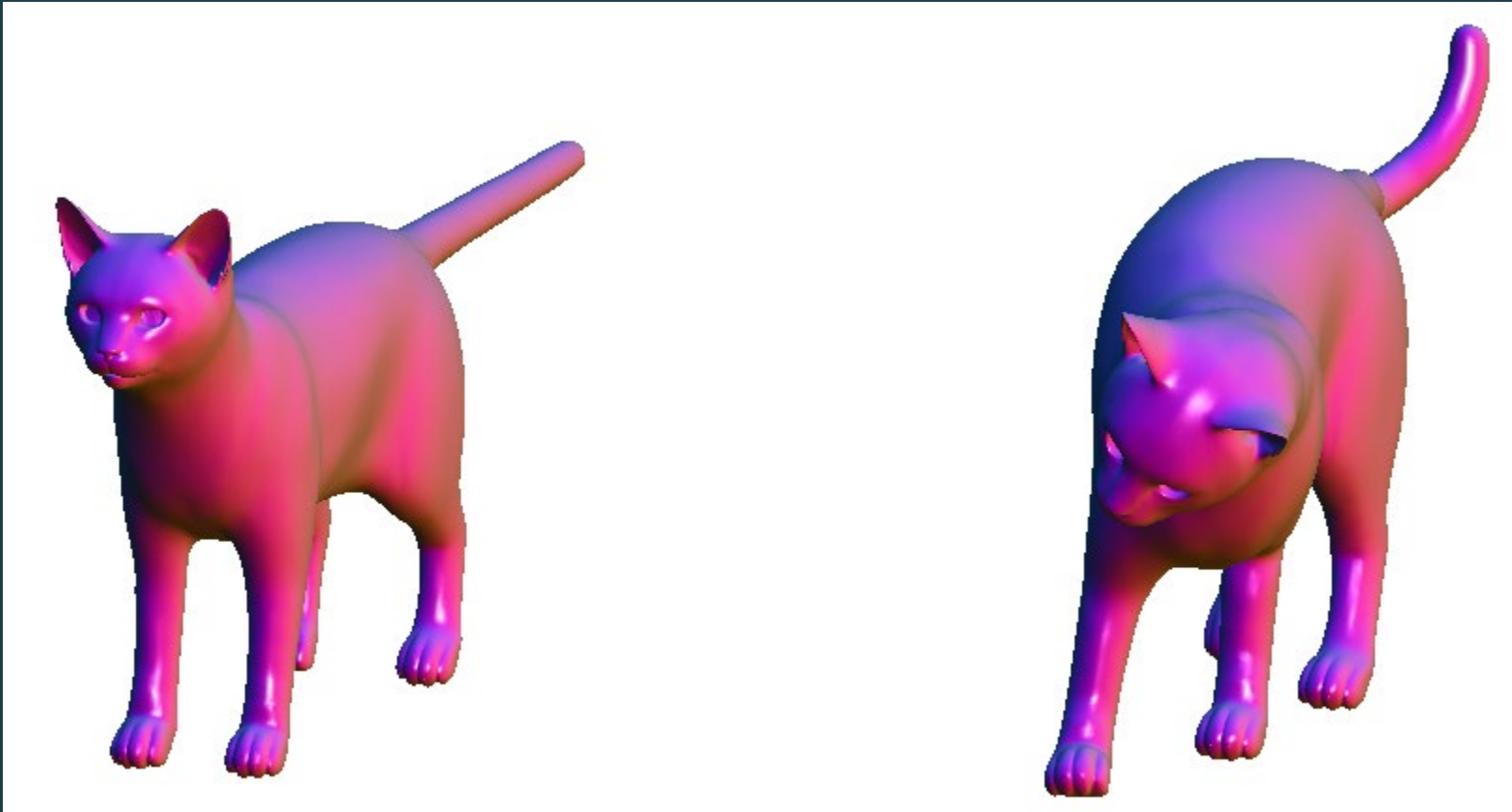


## Step 5 – Sphere Optical Flow





- Shapes having more non-isometric deformation



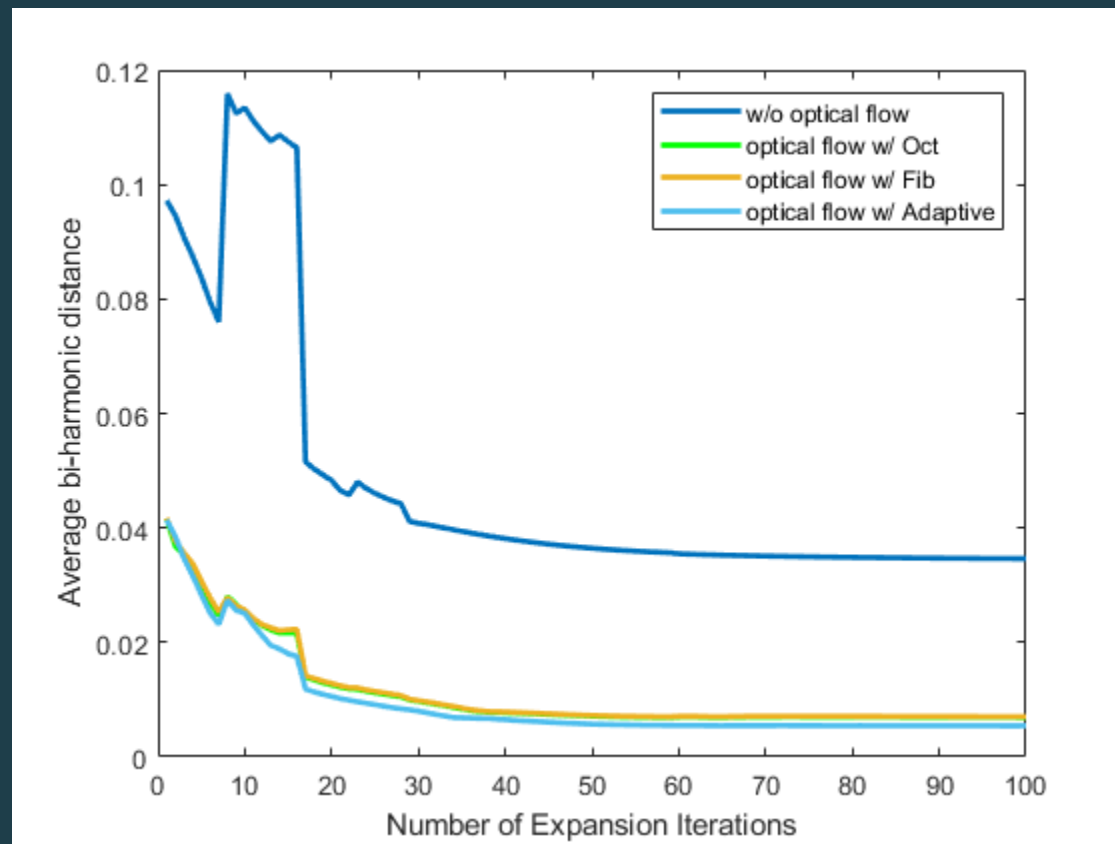
# Without Constraint



# With Constraint



# Comparison





*Feedback?*