



Symmetry Analysis and its Applications in Computer Graphics

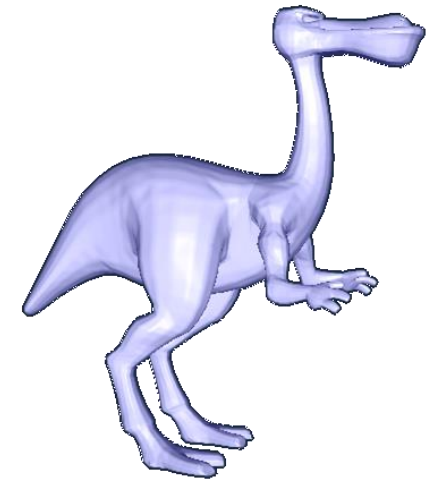
Thomas Funkhouser
Alex Golovinskiy, Misha Kazhdan,
Vladimir Kim, Josh Podolak,
and Szymon Rusinkiewicz

Princeton University

Motivation



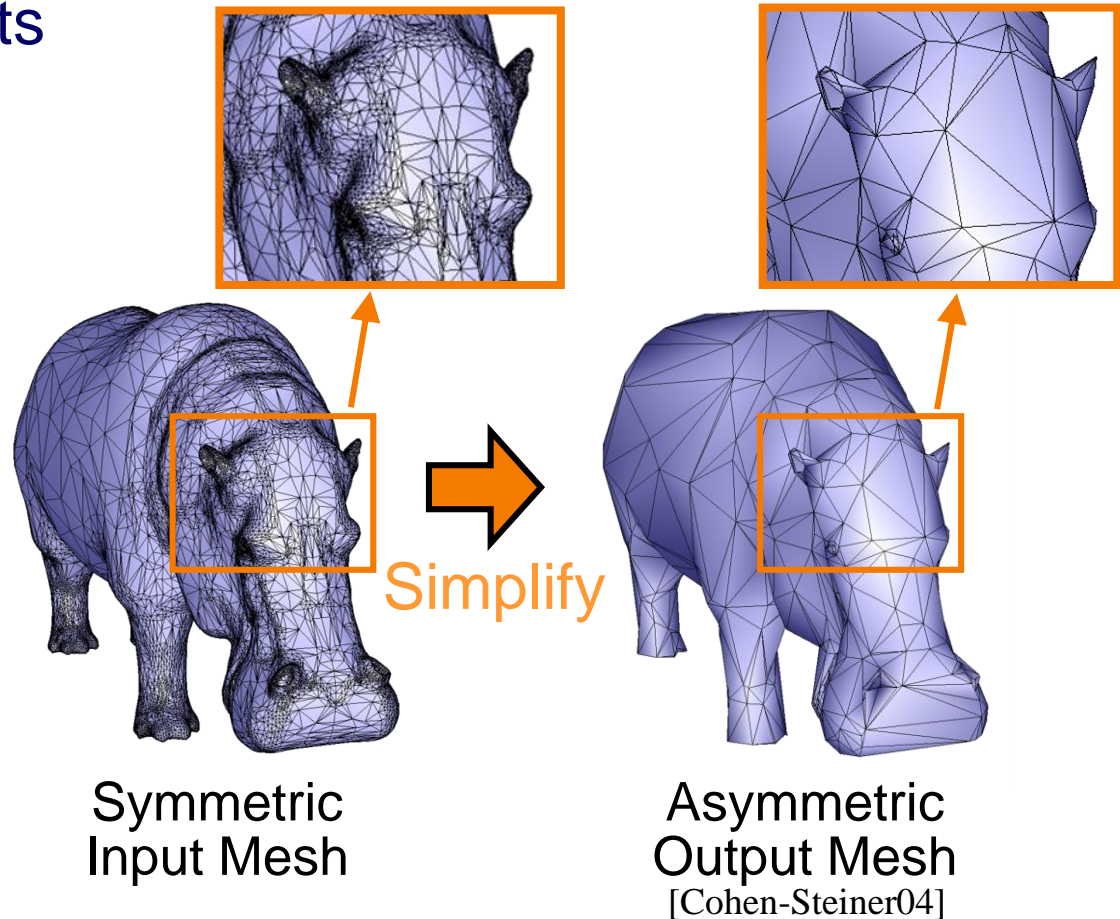
Symmetry is common in real-world objects



Problem

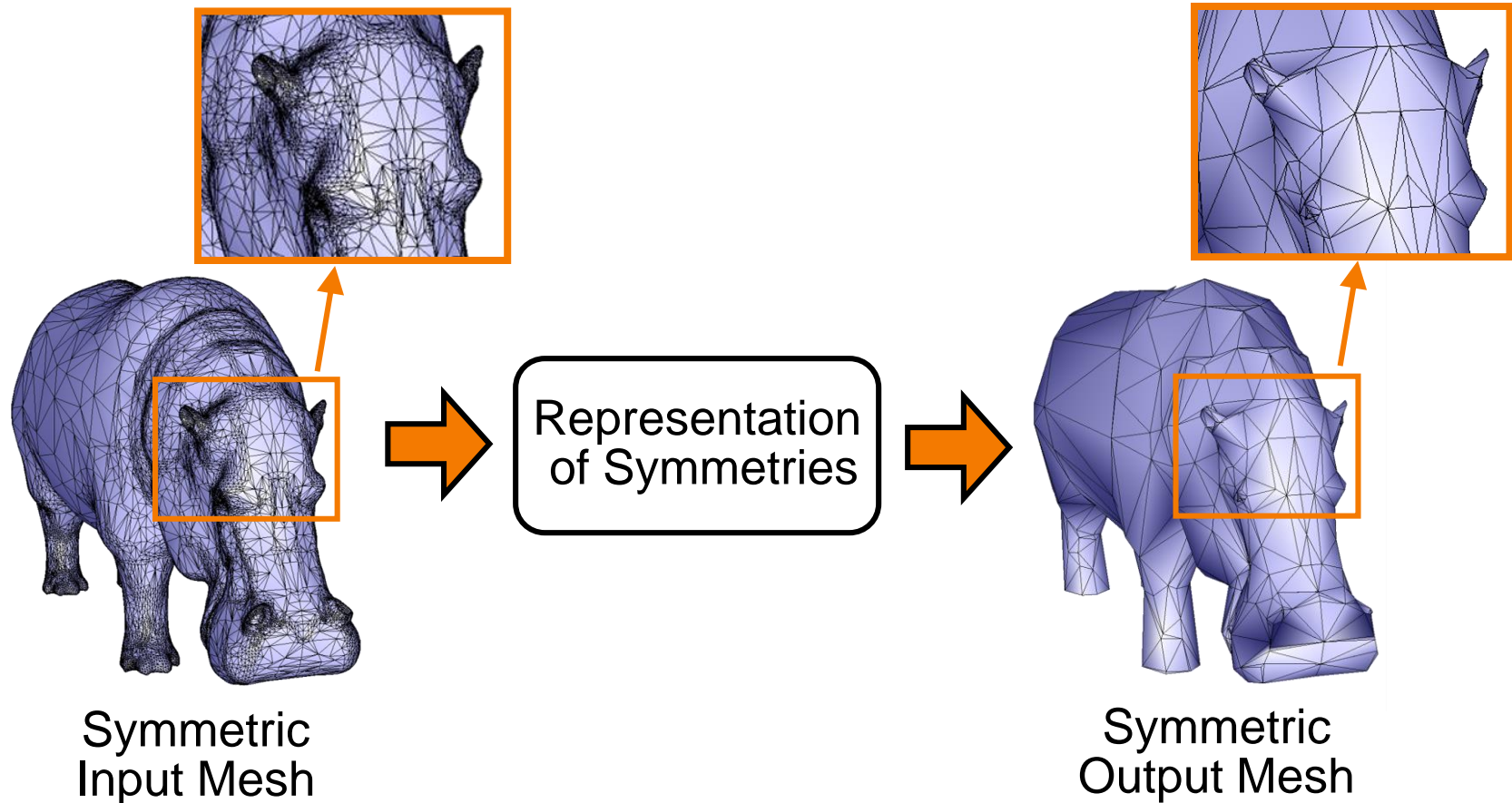
Most algorithms ignore (and sometimes destroy) symmetries when they process 3D models

- Rendering artifacts
- Simulation errors
- etc.



Goal

We need methods to detect, analyze, represent, and exploit symmetries in 3D models



Outline



Introduction

Background

Symmetry representations

- Symmetry descriptor
- Symmetry transform
- Principal symmetries

Applications

Conclusion

Outline



Introduction

Background ←

Symmetry representations

- Symmetry descriptor
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- Principal symmetries

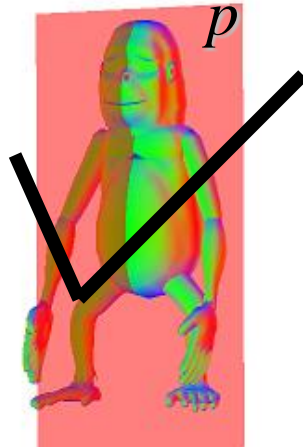
Applications

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Symmetry



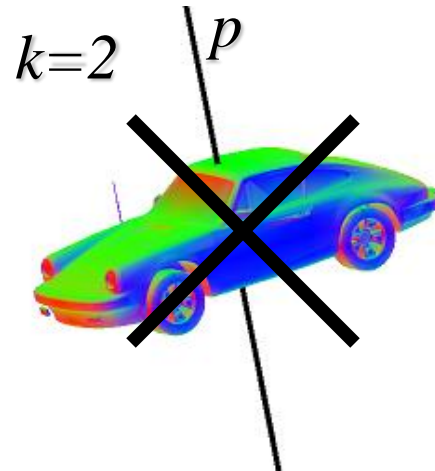
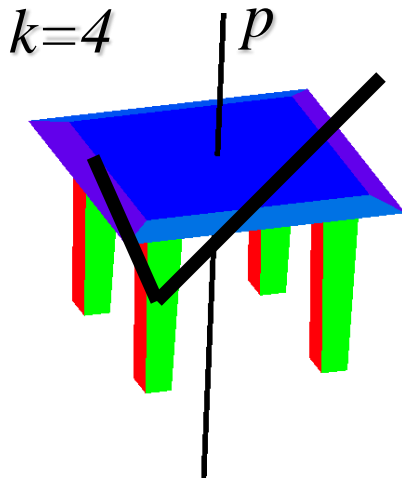
A shape has *reflective symmetry* w.r.t. some plane p if the reflection Ref_p through p fixes the collection



Symmetry



A shape has *rotational symmetry* of order k w.r.t. some axis p if the rotation Rot_p^k by an angle of $360^\circ/k$ about p fixes the collection.



Approximate Symmetry

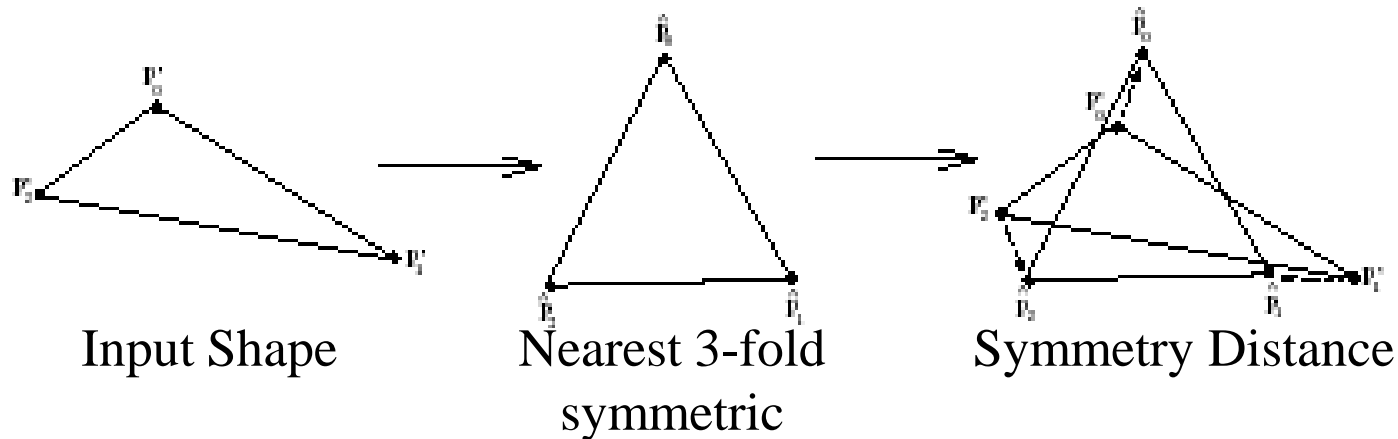


... but many shapes in computer graphics are not perfectly symmetric



Symmetry Distance

The *symmetry distance* of a boundary is the L^2 distance to the nearest symmetric boundary



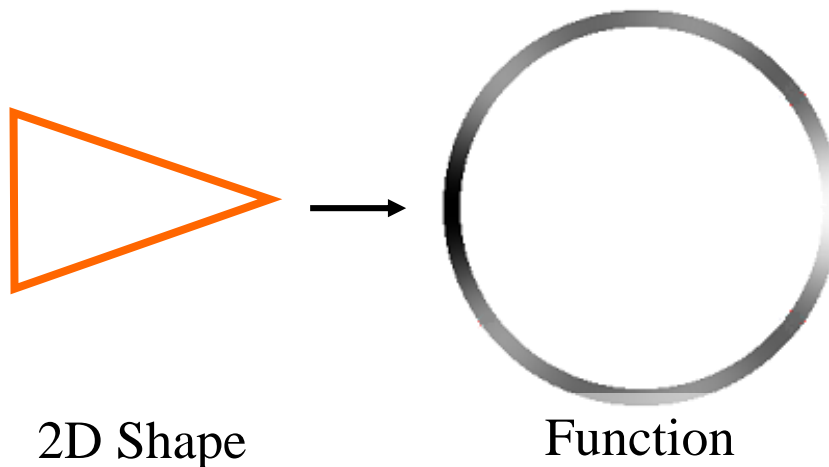
Requires point correspondences

IWVF, 1994. Zabrodsky et al.
IEEE, 1995. Zabrodsky et al.

Symmetry Distance



For circular functions, can replace distances between correspondences with correlation

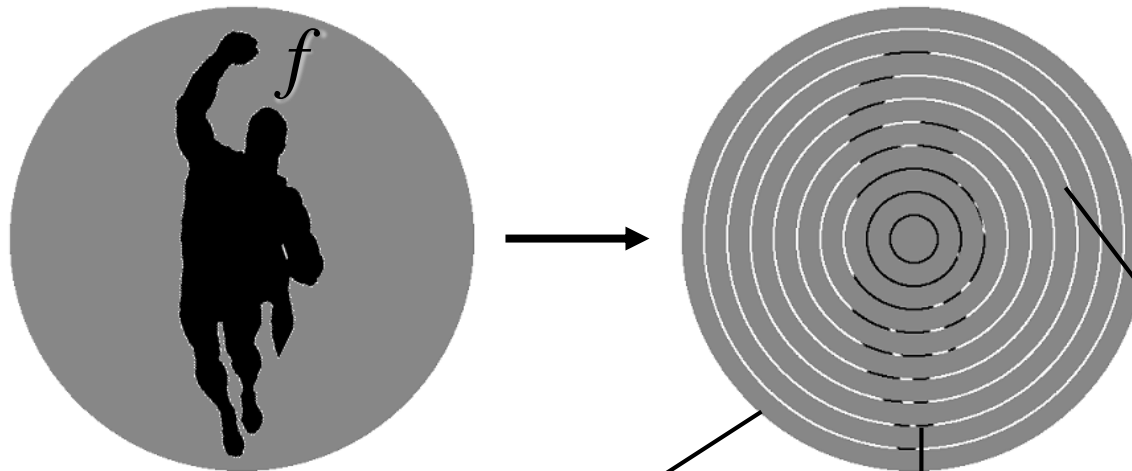


$$\text{Sym}(SD, \alpha) = \int_0^{2\pi} SD(t - \alpha)SD(t)dt$$

PRL, 1995. Sun
RTI, 1999. Sun et al.

Symmetry Distance

For images, can consider circular restrictions and compute correlations efficiently with Fourier transform



$$Sym(f, \gamma) = \underbrace{r_1}_{\text{scale factor}} Sym(\cdot, \gamma) + \underbrace{r_2}_{\text{scale factor}} Sym(\cdot, \gamma) + \underbrace{r_3}_{\text{scale factor}} Sym(\cdot, \gamma) + \dots$$

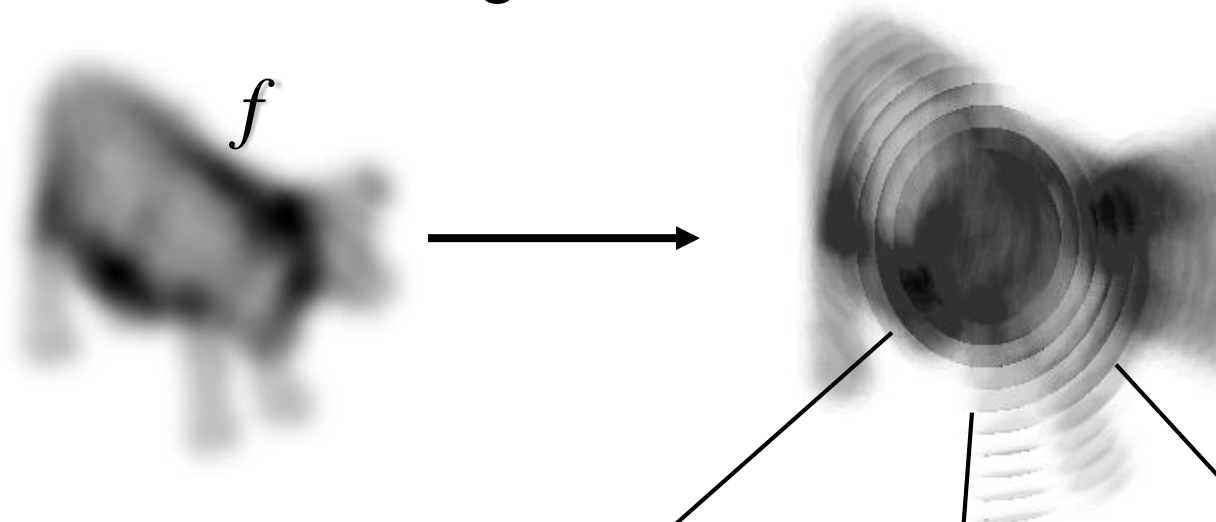
PRL, 1995. Sun

RTI, 1999. Sun et al.

Symmetry Distance



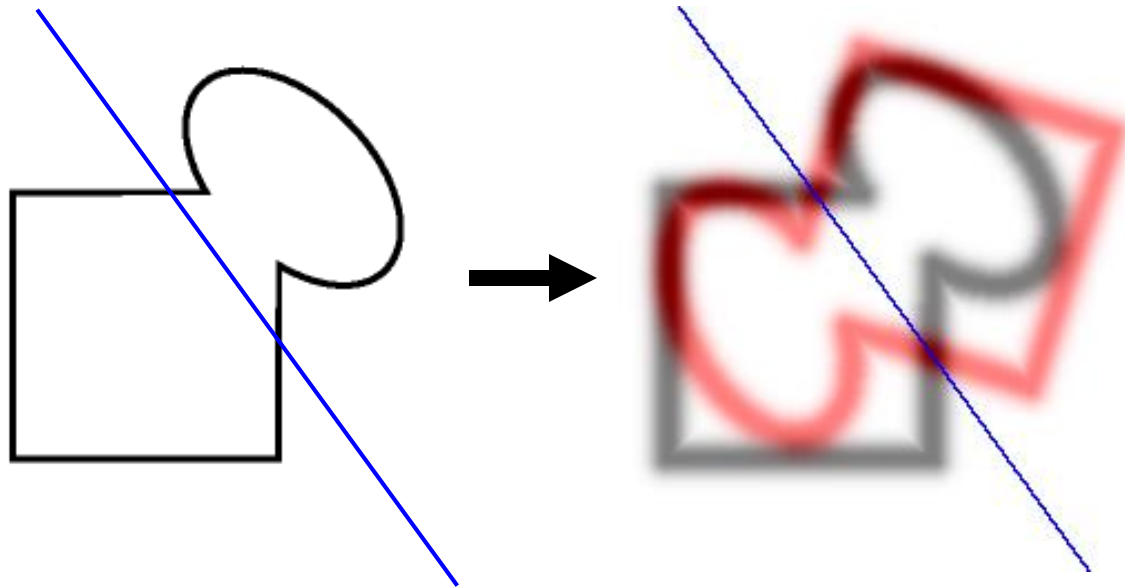
For 3D functions, can consider spherical restrictions and compute correlation efficiently with spherical harmonics and Wigner-D⁻¹ transform



$$\text{Sym}(f, R) = \underbrace{r_1}_{\text{scale factor}} \text{Sym}(\cdot, R) + \underbrace{r_2}_{\text{scale factor}} \text{Sym}(\cdot, R) + \underbrace{r_3}_{\text{scale factor}} \text{Sym}(\cdot, R) + \dots$$

Symmetry Distance

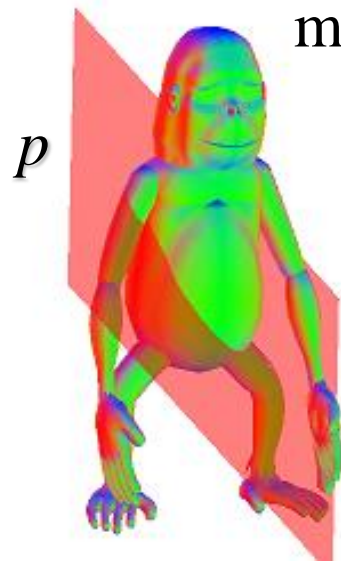
For boundaries, can approximate symmetry distance by converting to a function and then computing correlation



Symmetry Distance



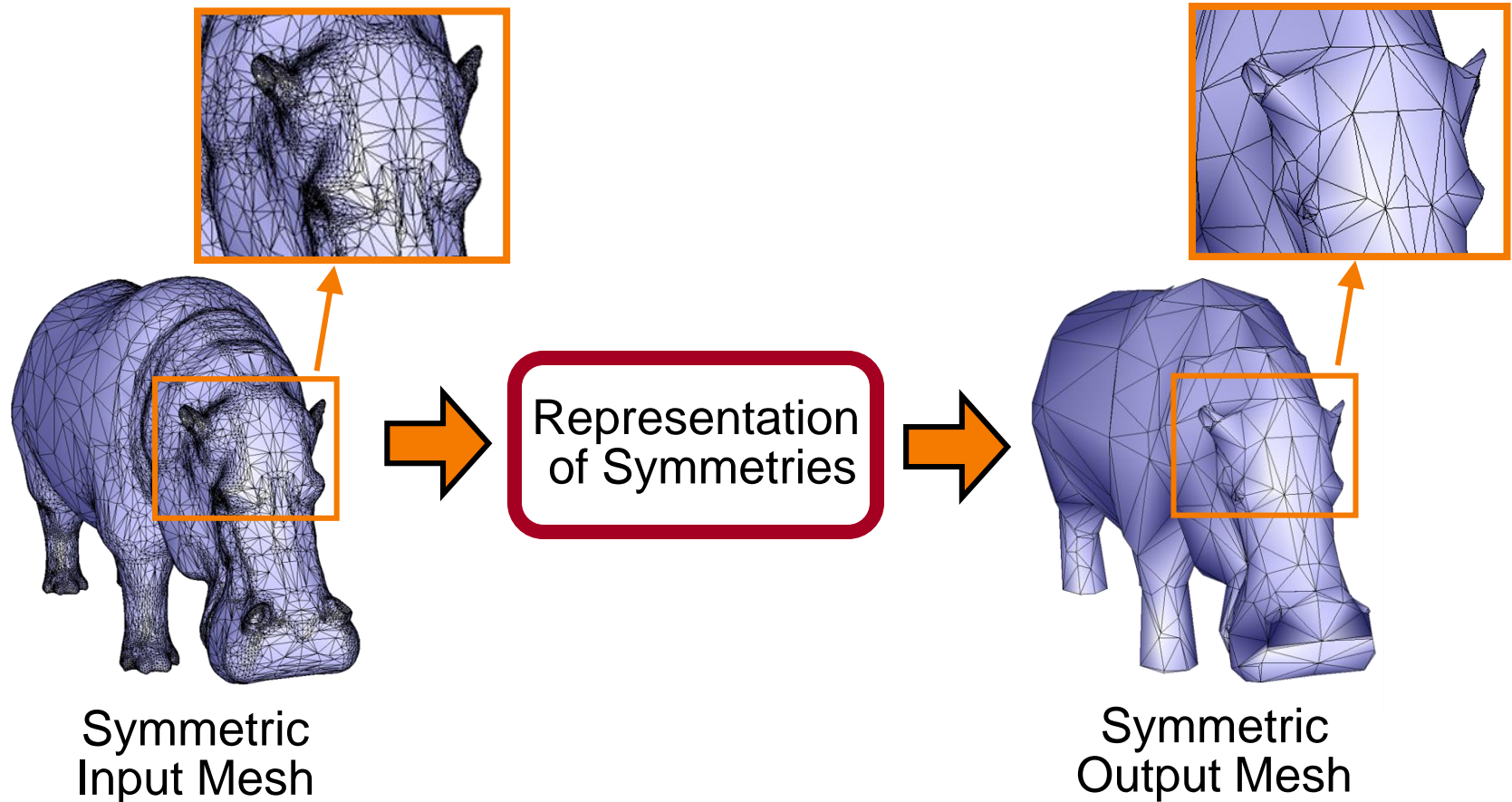
Can use this method to compute symmetry distance for any 3D mesh (without finding correspondences)



$$\text{Sym}(m,p) = 0.3$$

Back to the Goal ...

We need methods to detect, analyze, **represent**, and exploit symmetries in 3D models



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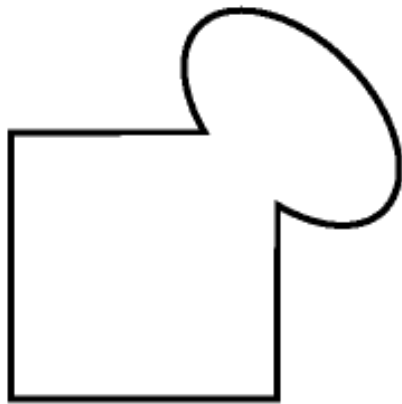
Applications

Conclusion

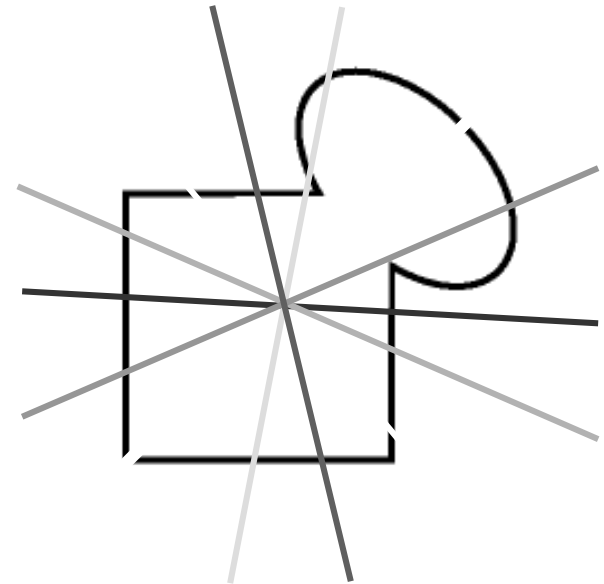
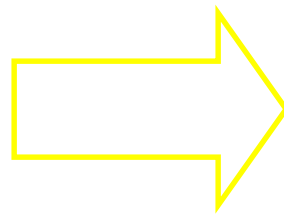
Symmetry Descriptor



Measure the symmetry of an object with respect to every transformation through its center



Input Shape



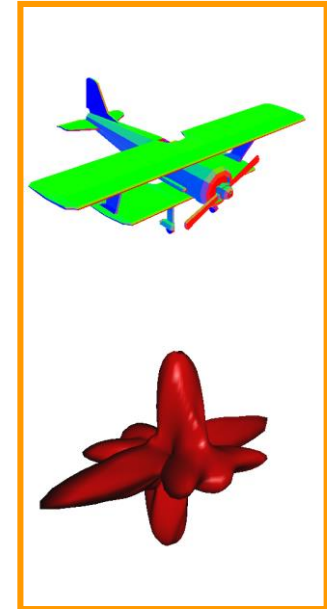
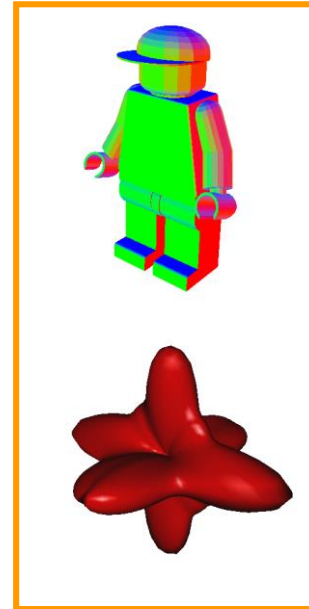
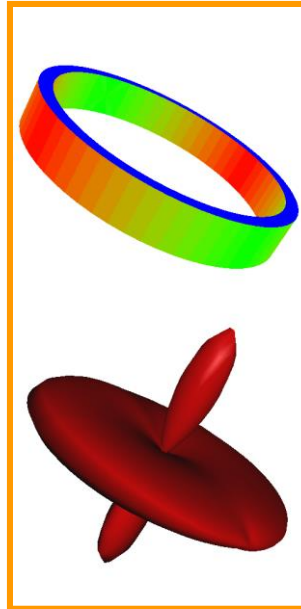
**Planar Reflective
Symmetry Descriptor**

Symmetry Descriptor



Planar reflective symmetry descriptors:

Input Mesh

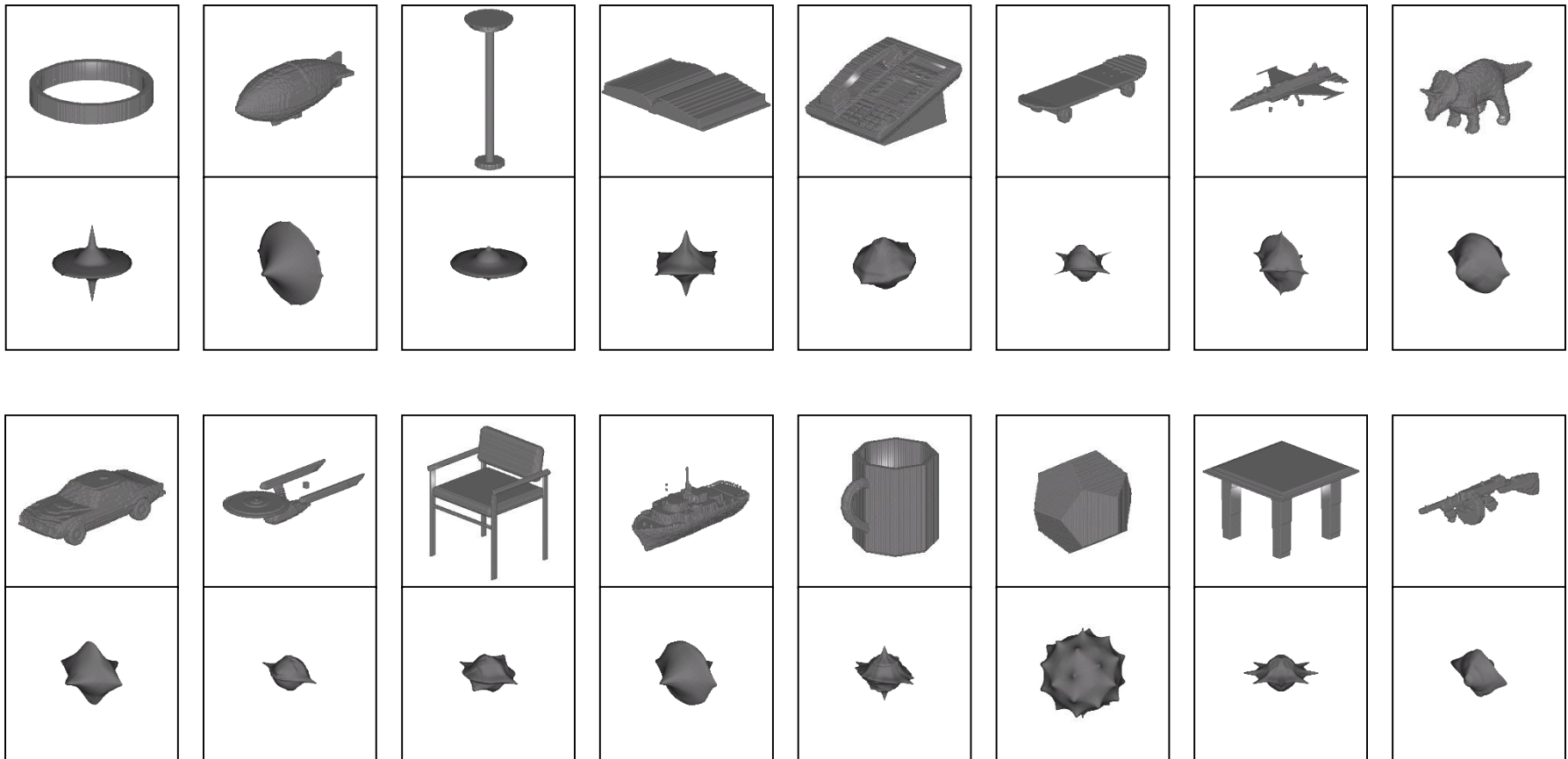


Planar Reflective
Symmetry Descriptor

Symmetry Descriptor



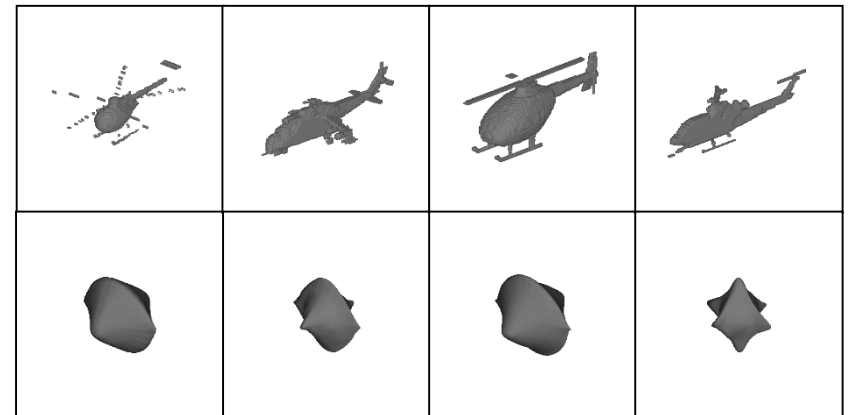
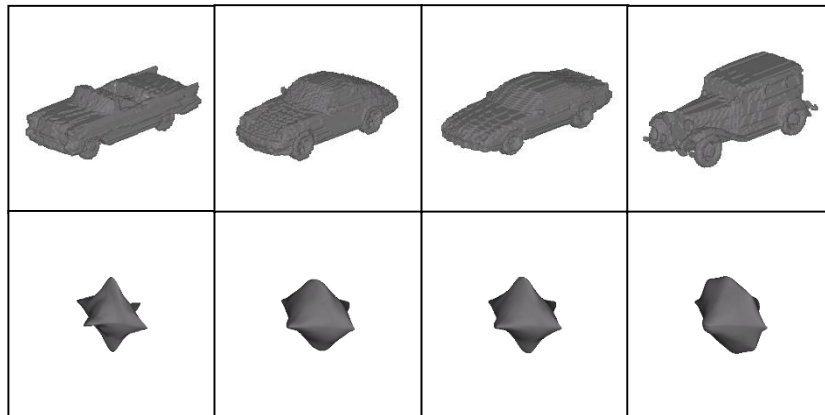
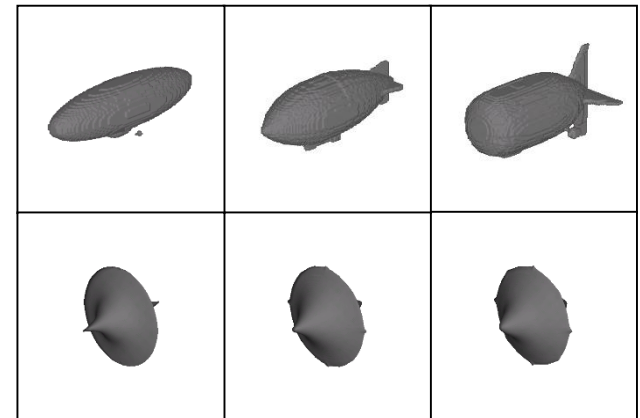
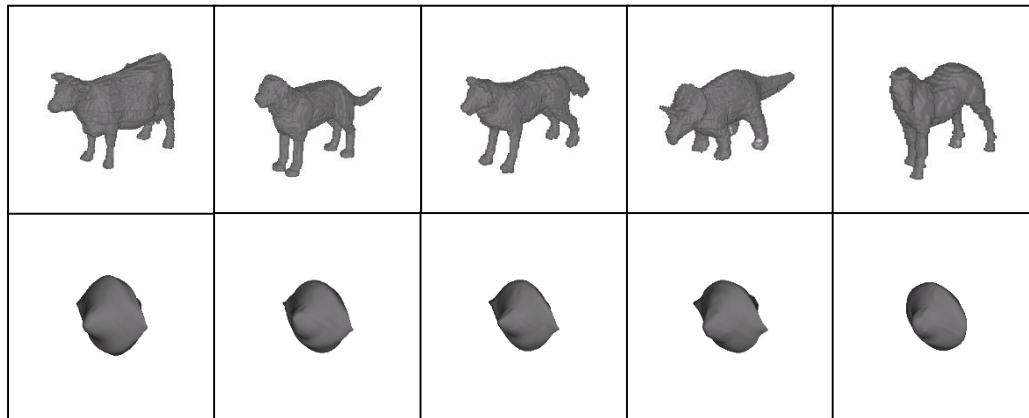
Planar reflective symmetry descriptors:



Symmetry Descriptor



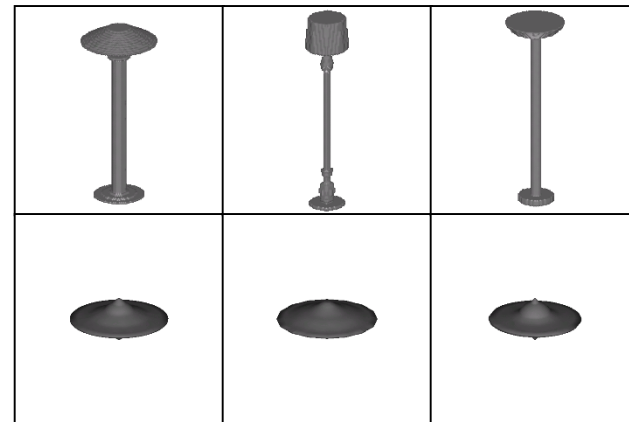
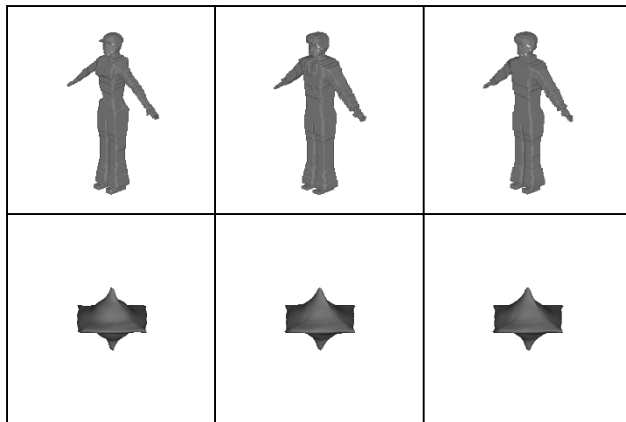
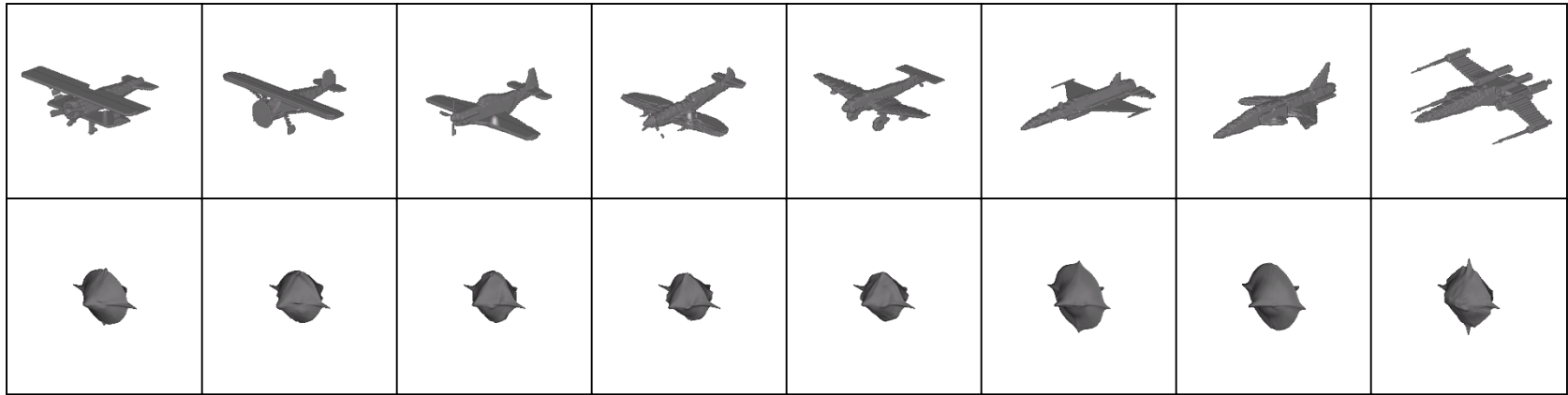
Planar reflective symmetry descriptors:



Symmetry Descriptor



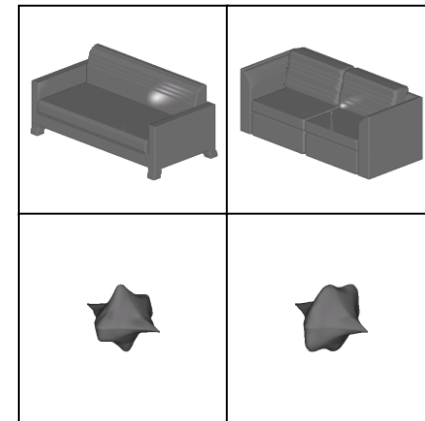
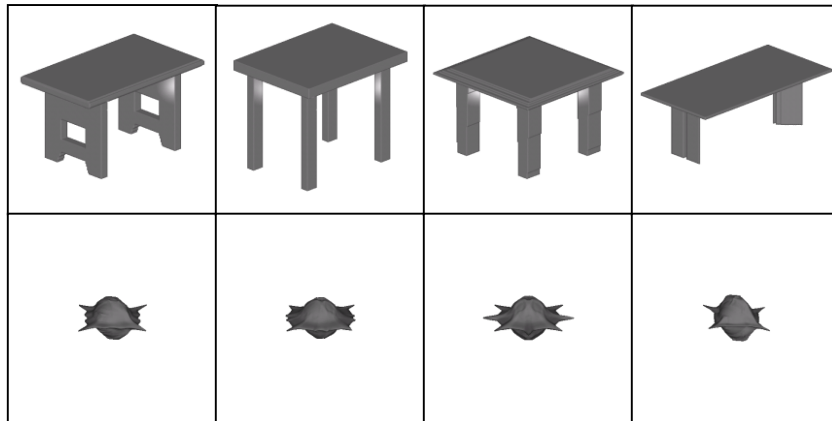
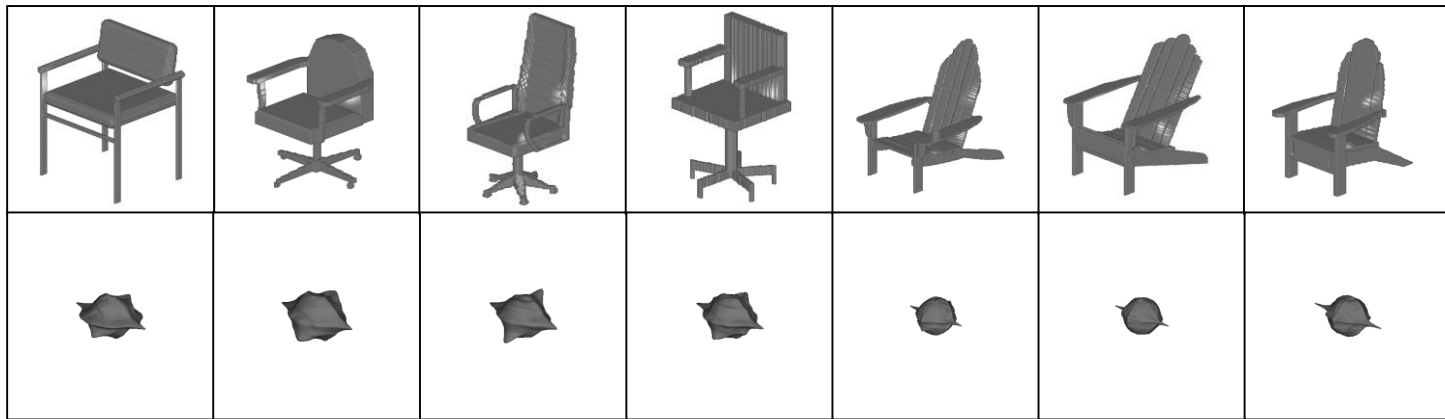
Planar reflective symmetry descriptors:



Symmetry Descriptor





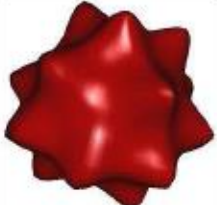


















Planar reflective symmetry descriptors:



Symmetry Descriptor



Rotational symmetry descriptors:

						
						
						
	$k=-4$	$k=-2$ (reflective)	$k=2$	$k=3$	$k=4$	$k=\infty$ (axial)

Symmetry Descriptor



Rotational symmetry descriptors:

	$k=-4$	$k=-2$ (reflective)	$k=2$	$k=3$	$k=4$	$k=\infty$ (axial)

Symmetry Descriptor



Properties:

- Canonical parameterization
 - Parameterized over the (projective) sphere
- Insensitive to noise:
 - Integration scales down high frequency Fourier coefficients
- Global:
 - For functions f and g , and any reflection r :

$$|\text{Sym}(f, r) - \text{Sym}(g, r)| \leq \|f - g\|_2$$

Outline



Introduction

Background

Symmetry representations

- Symmetry descriptor
- **Symmetry transform** ←
- Principal symmetries

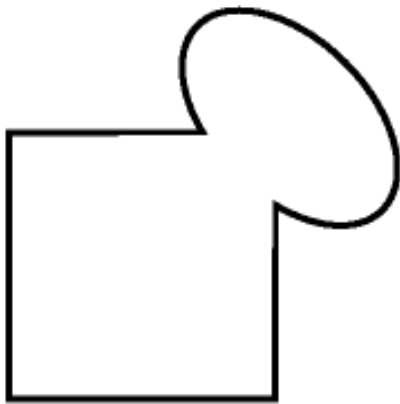
Applications

Conclusion

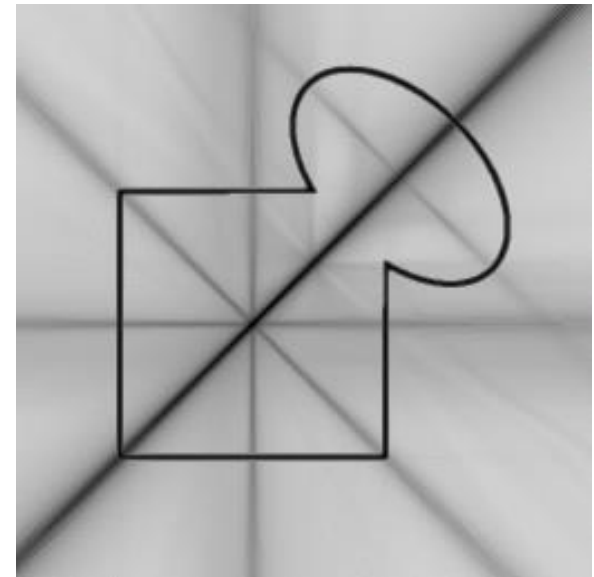
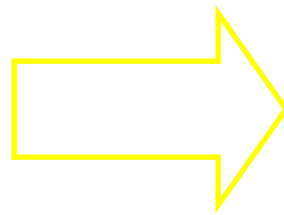
Symmetry Transform



Measure the symmetry of an object with respect to all transformations



Input Shape

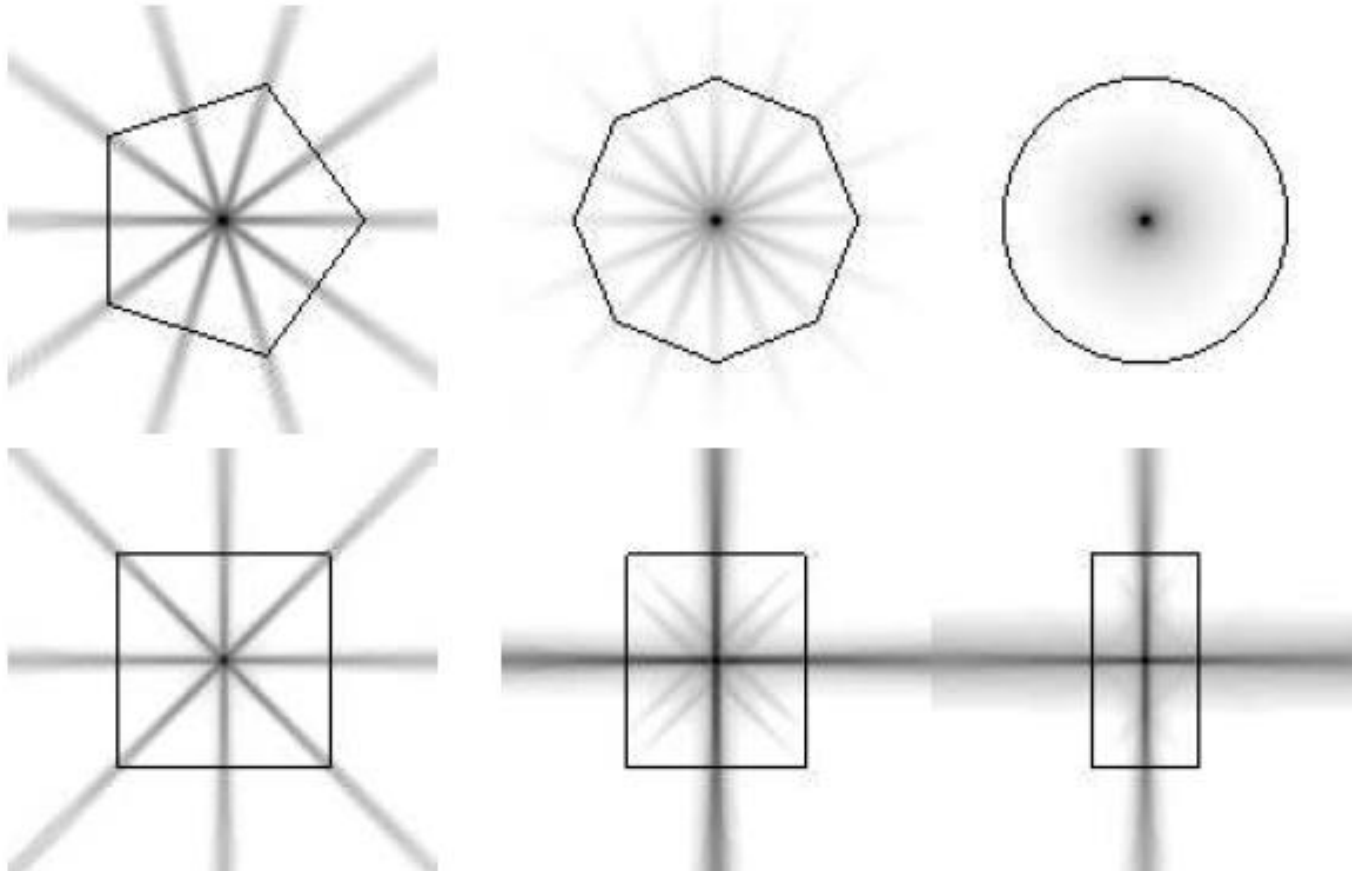


Planar Reflective Symmetry Transform

Symmetry Transform



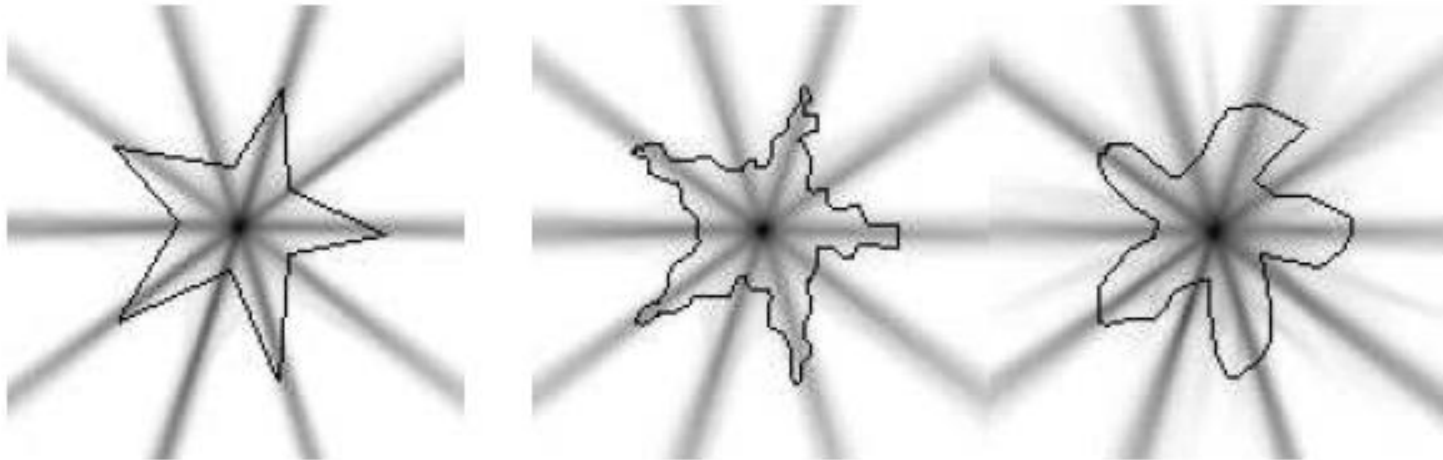
Planar reflective symmetry transform:



Symmetry Transform



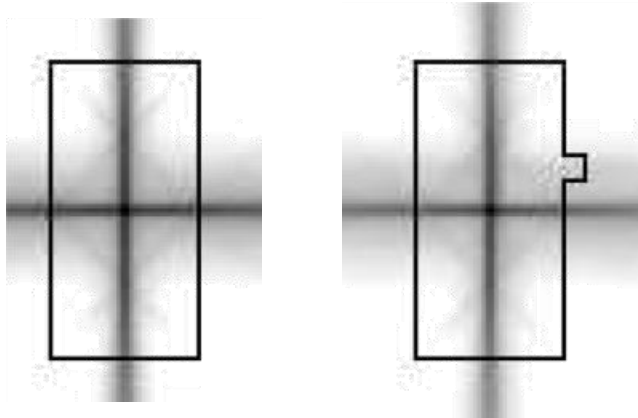
Stability with noise:



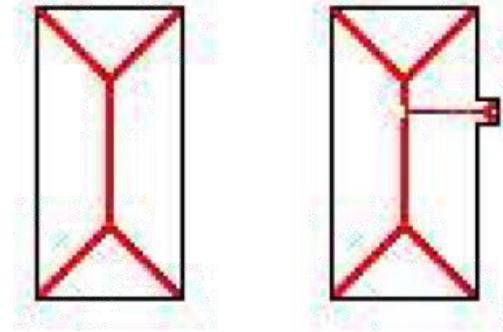
Symmetry Transform



Stability with small extra features:



Planar Reflective
Symmetry Transform

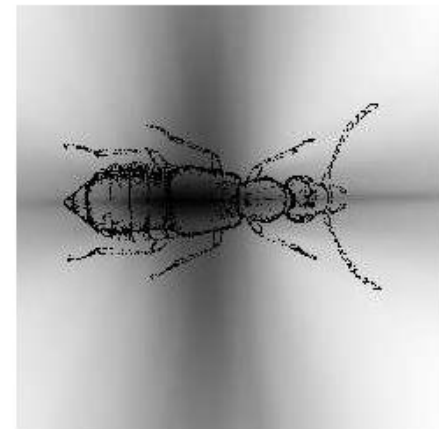
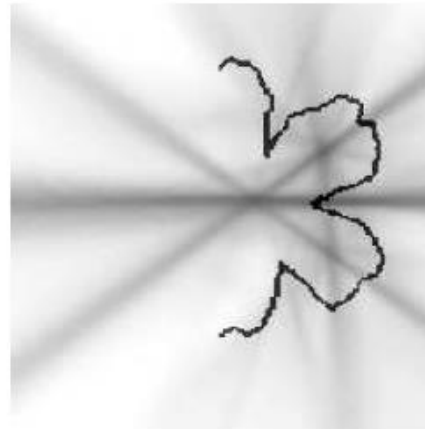
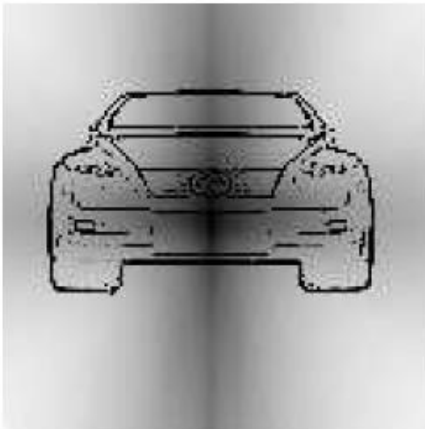
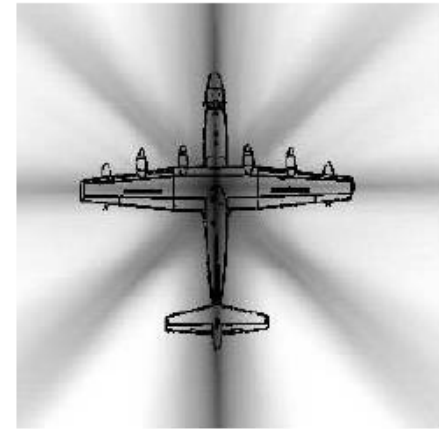
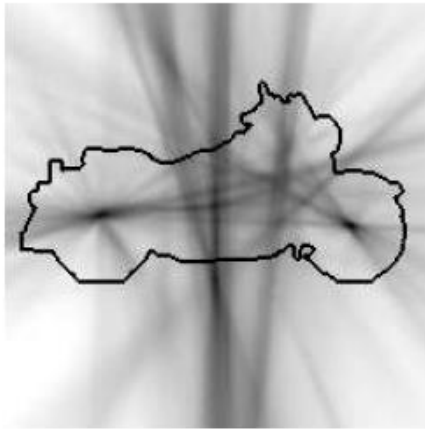


Medial Axis Transform

Symmetry Transform



Highlights large symmetric features of shape:



Symmetry Transform



Computation:

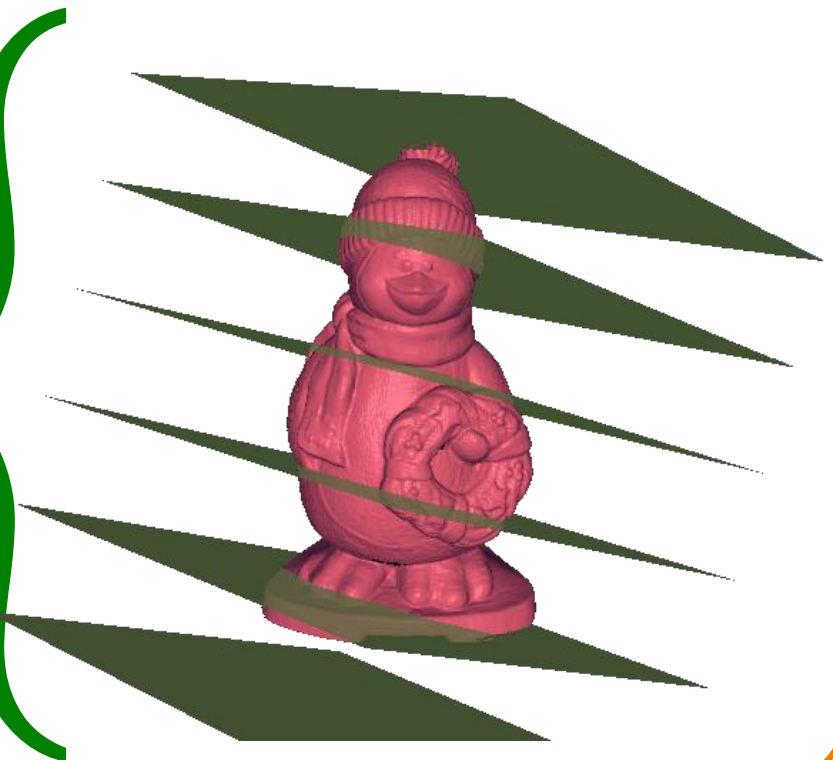
➤ Brute Force: $O(n^6)$

$O(n^3)$ planes

$X = O(n^6)$

$O(n^3)$ correlation

n planes



Symmetry Transform



Computation:

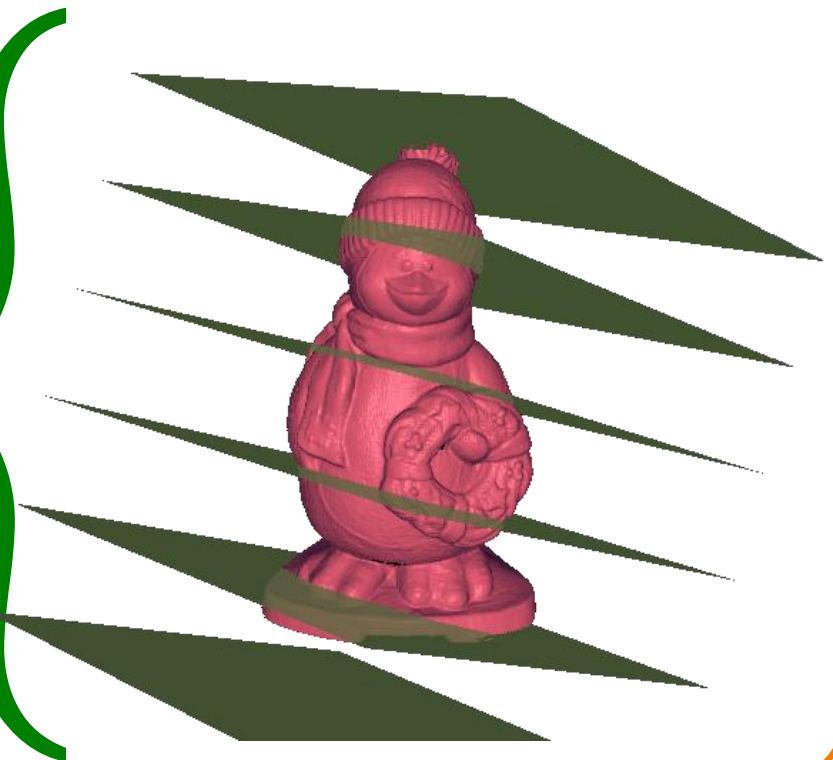
- Brute Force: $O(n^6)$
- Convolution: $O(n^5 \log n)$

$O(n^2)$ normal directions

$X = O(n^5 \log n)$

$O(n^3 \log n)$ per direction

n planes

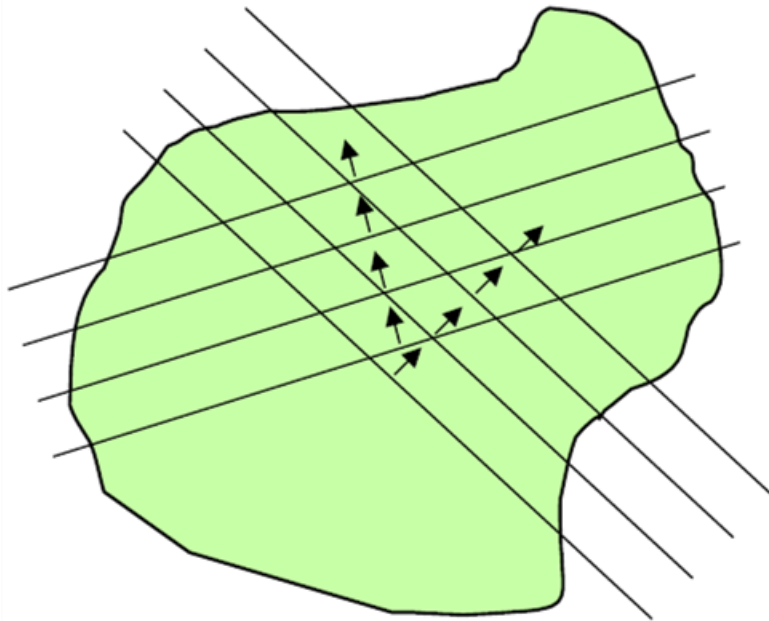


Symmetry Transform

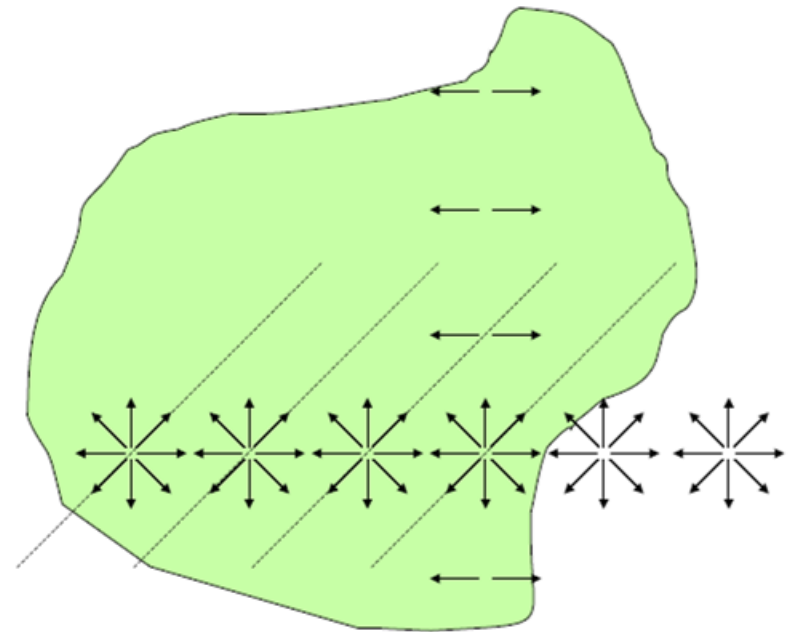


Computation:

- Brute Force: $O(n^6)$
- Convolution: $O(n^5 \log n)$



Translate planes



Rotate planes

Symmetry Transform

Computation:

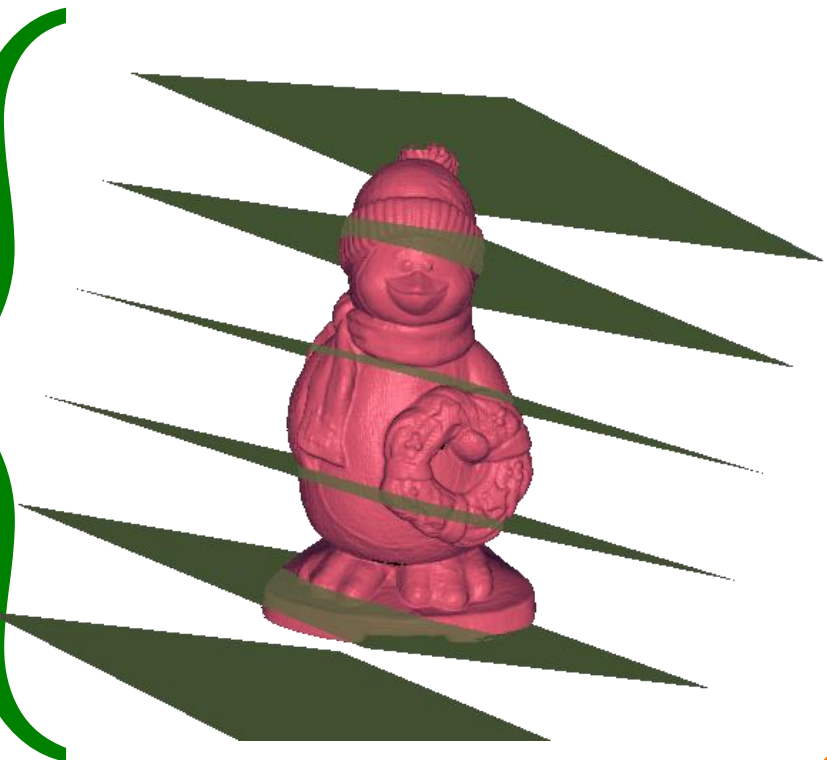
- Brute Force: $O(n^6)$
- Convolution: $O(n^5 \log n)$
- Monte Carlo: $O(n^4)$ for surfaces

$O(n^2)$ surface points

$X = O(n^4)$

$O(n^2)$ pairs

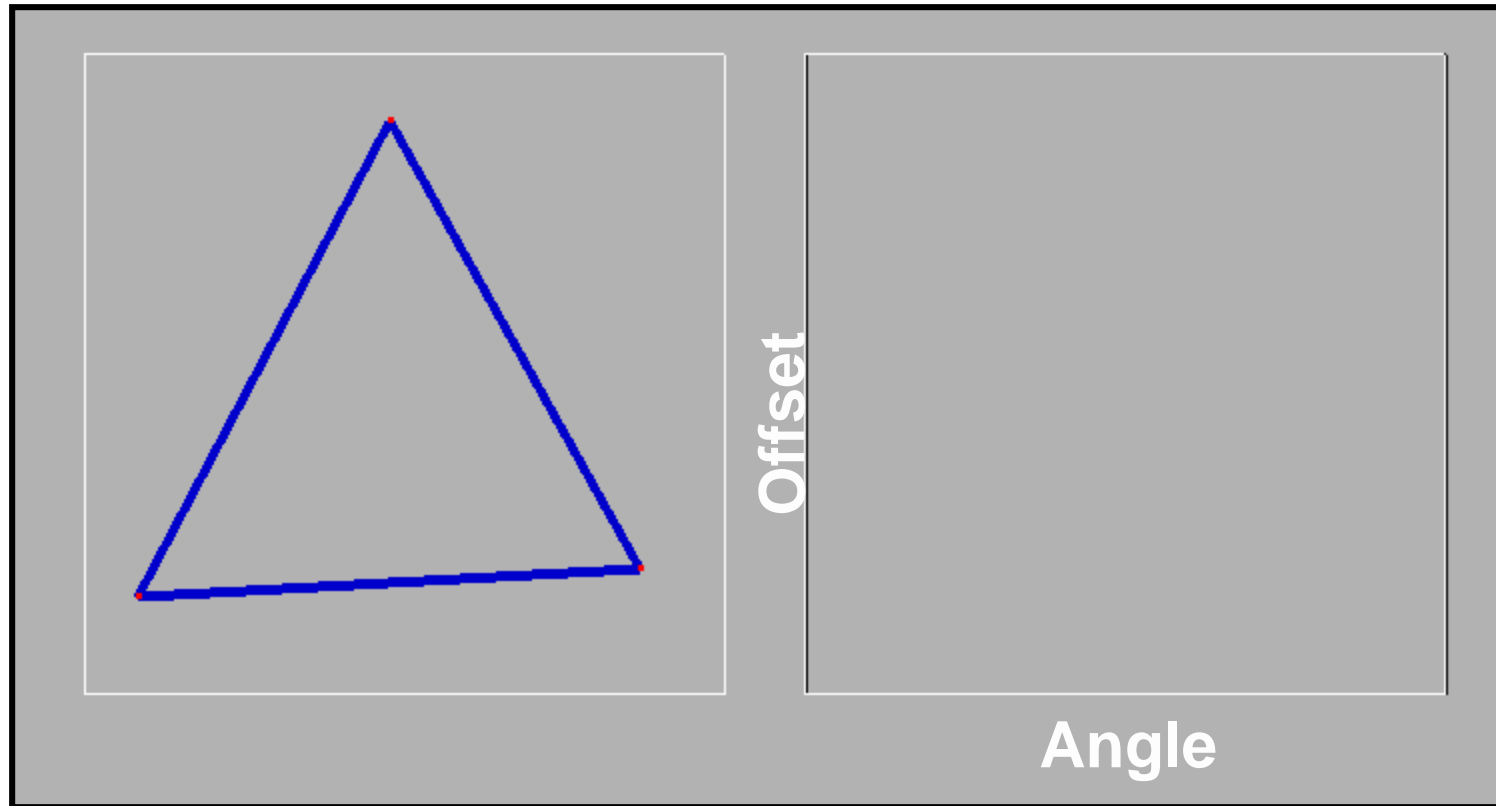
n planes



Symmetry Transform



Monte Carlo algorithm:



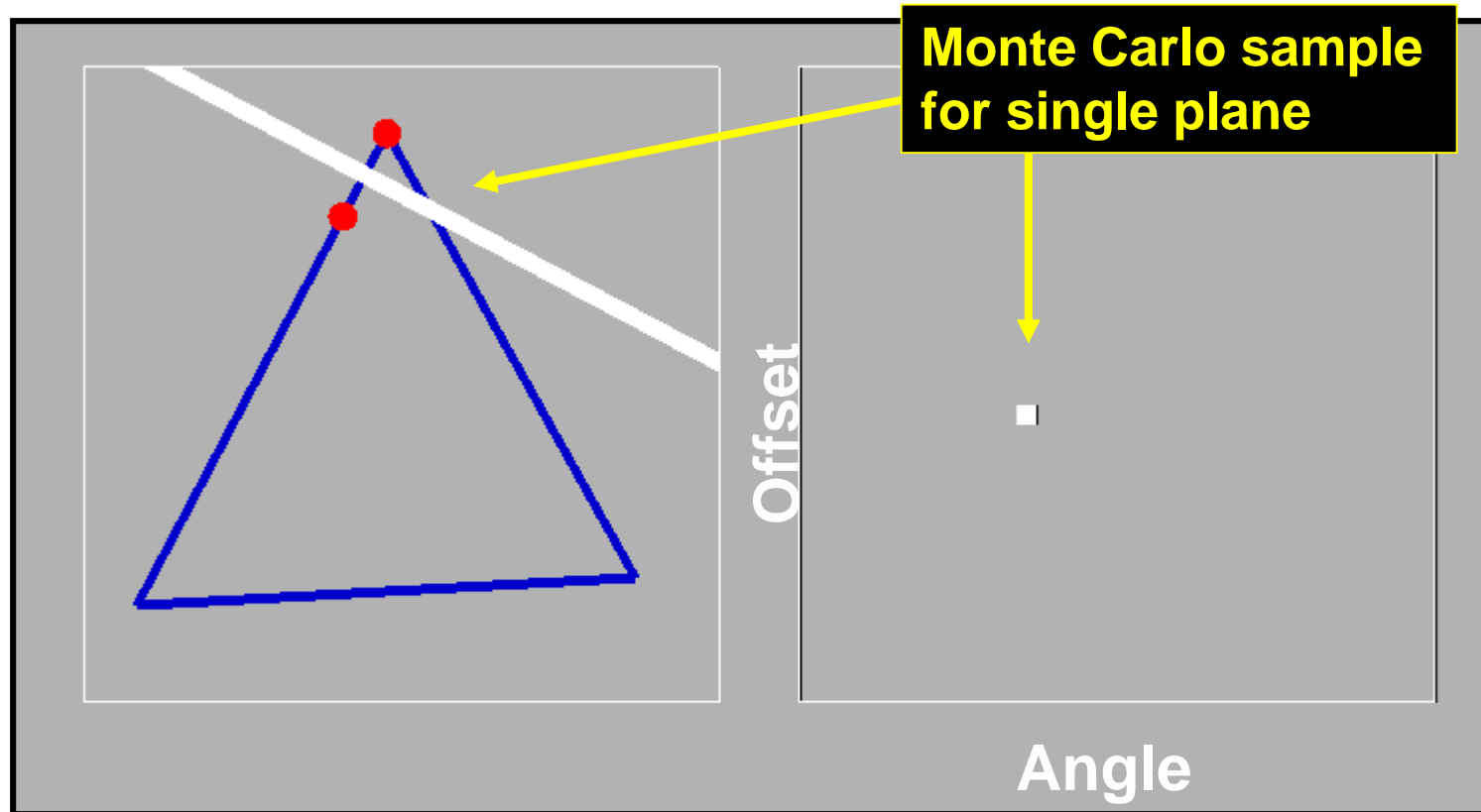
Input Model

Symmetry Transform

Symmetry Transform



Monte Carlo algorithm:



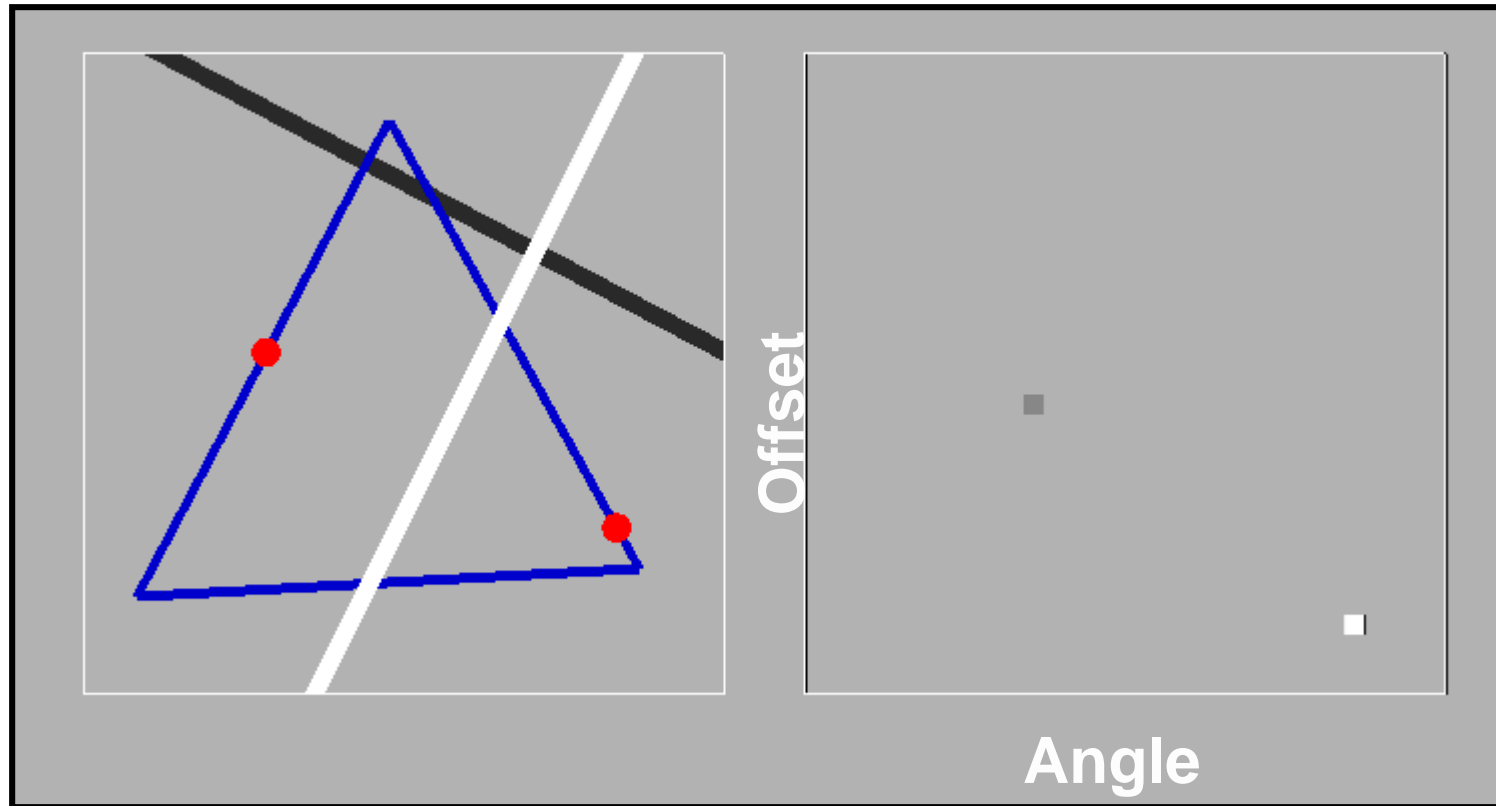
Input Model

Symmetry Transform

Symmetry Transform



Monte Carlo algorithm:



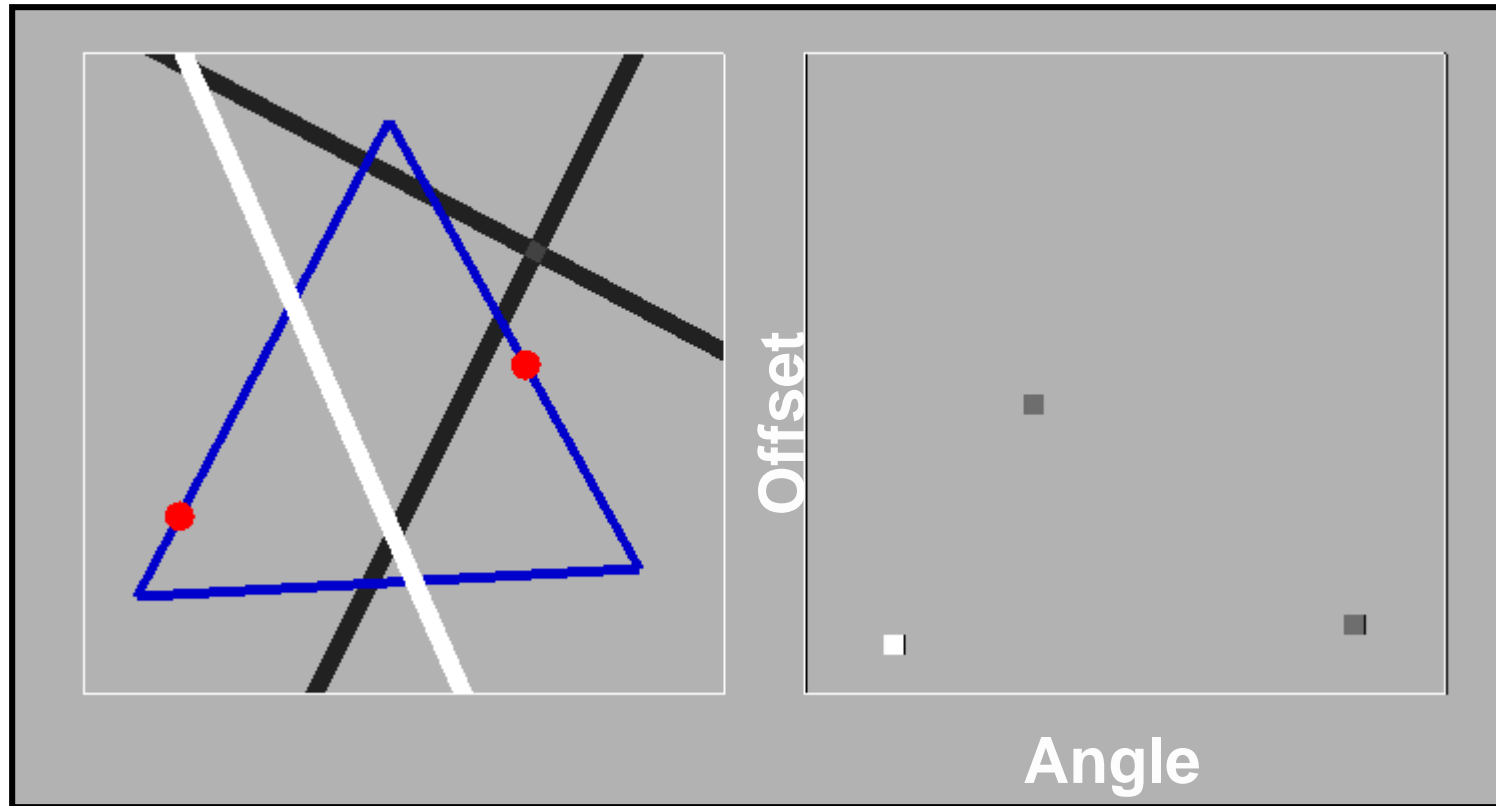
Input Model

Symmetry Transform

Symmetry Transform



Monte Carlo algorithm:



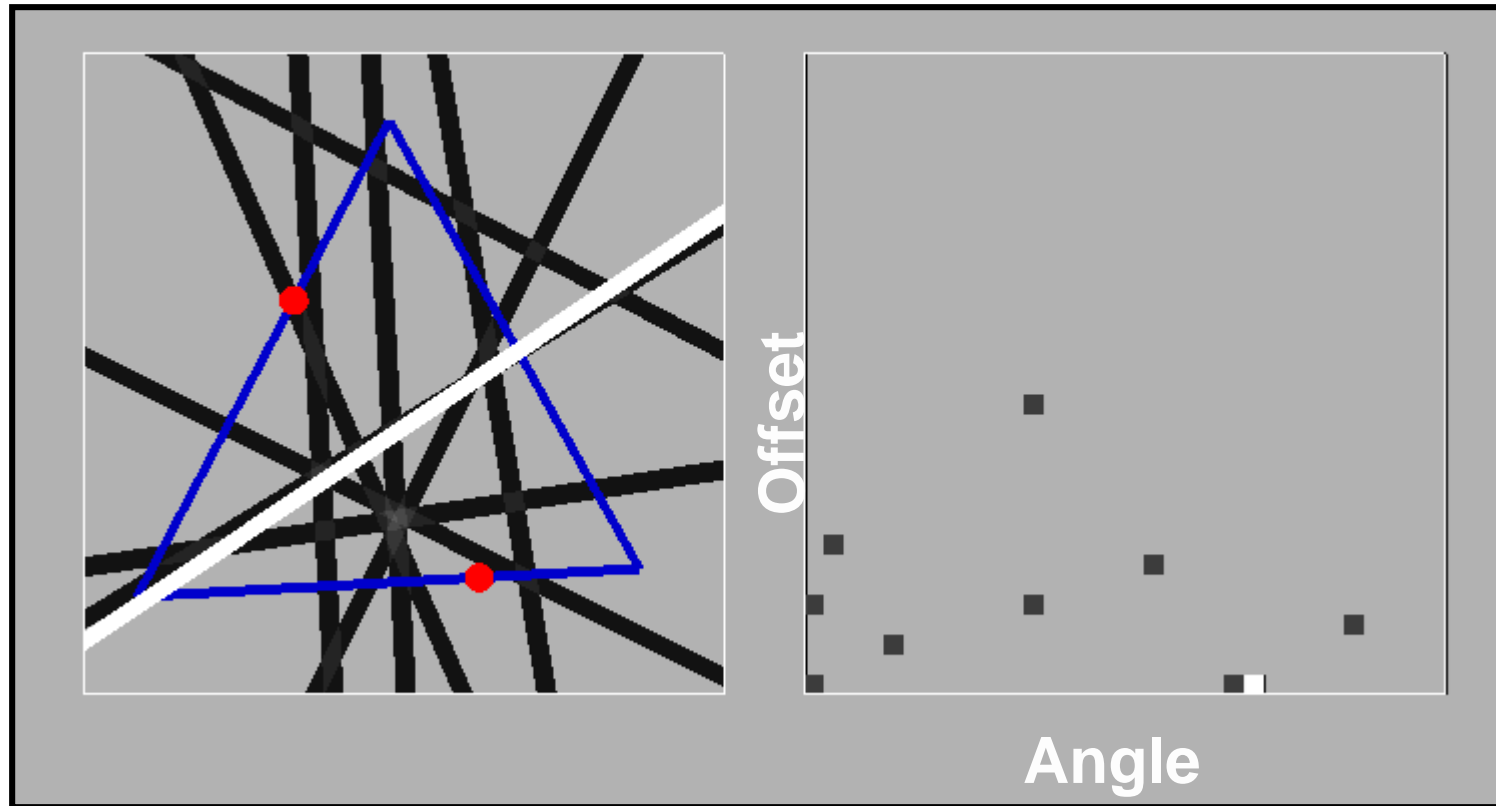
Input Model

Symmetry Transform

Symmetry Transform



Monte Carlo algorithm:



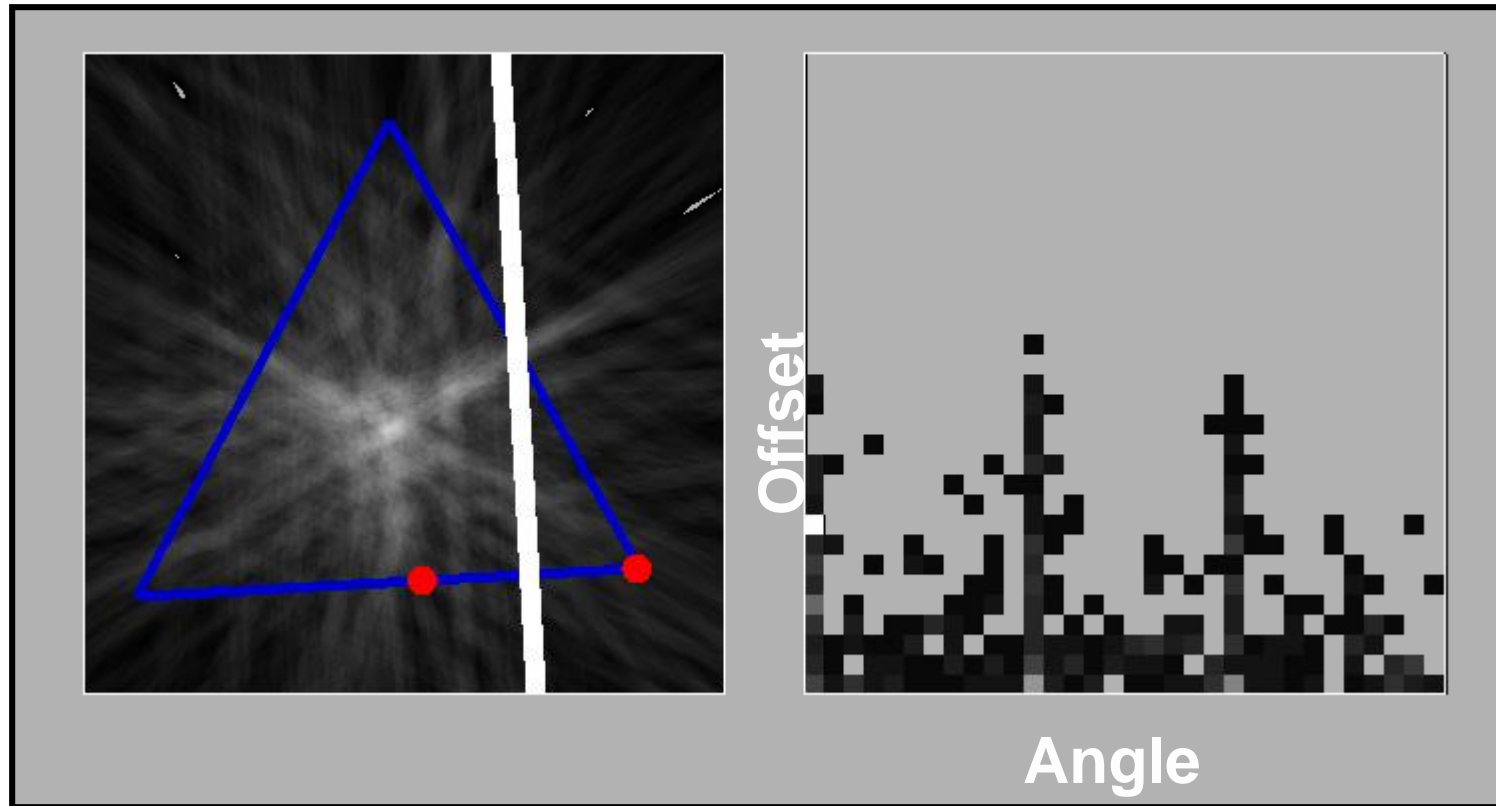
Input Model

Symmetry Transform

Symmetry Transform



Monte Carlo algorithm:



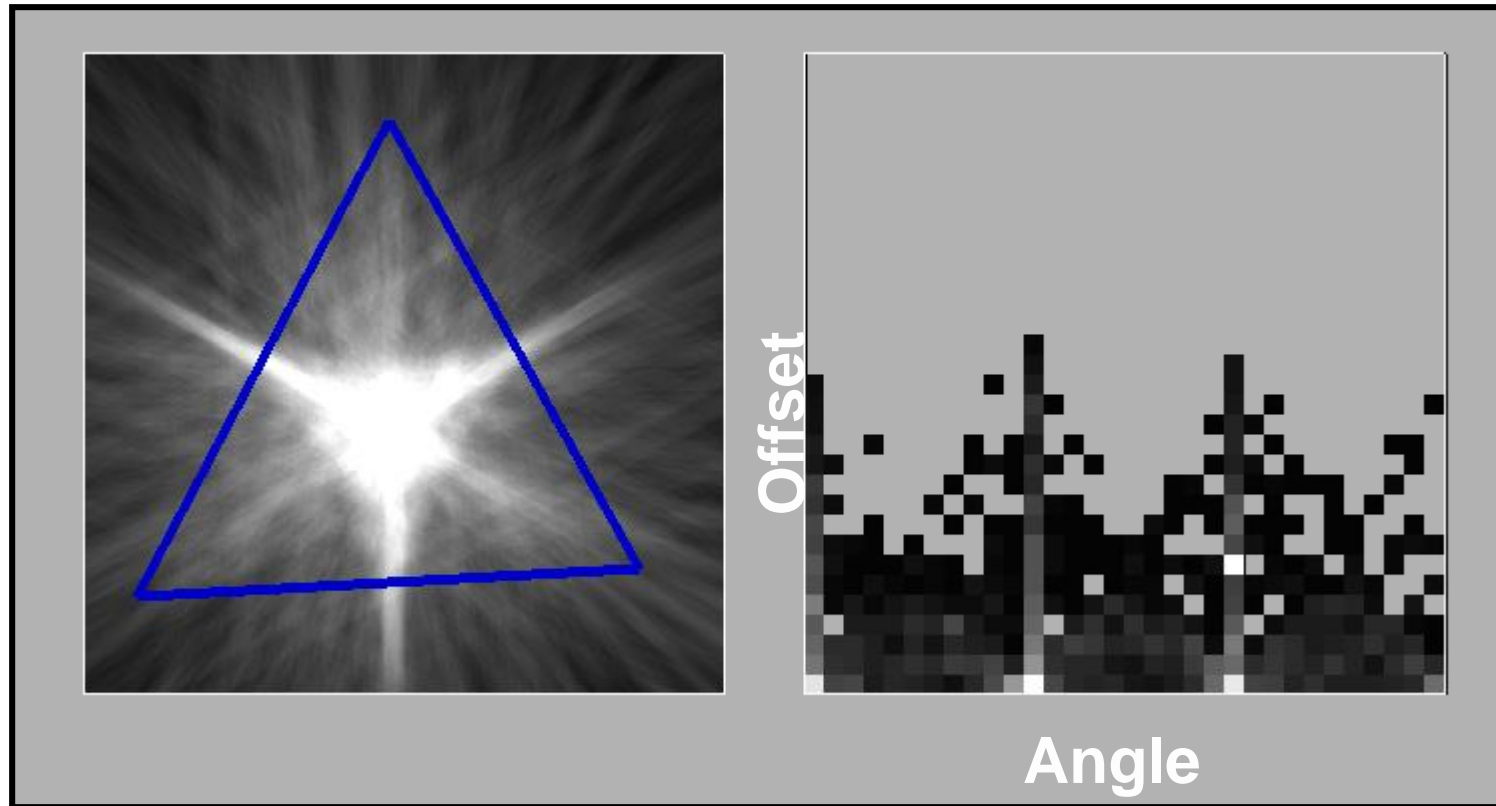
Input Model

Symmetry Transform

Symmetry Transform



Monte Carlo algorithm:



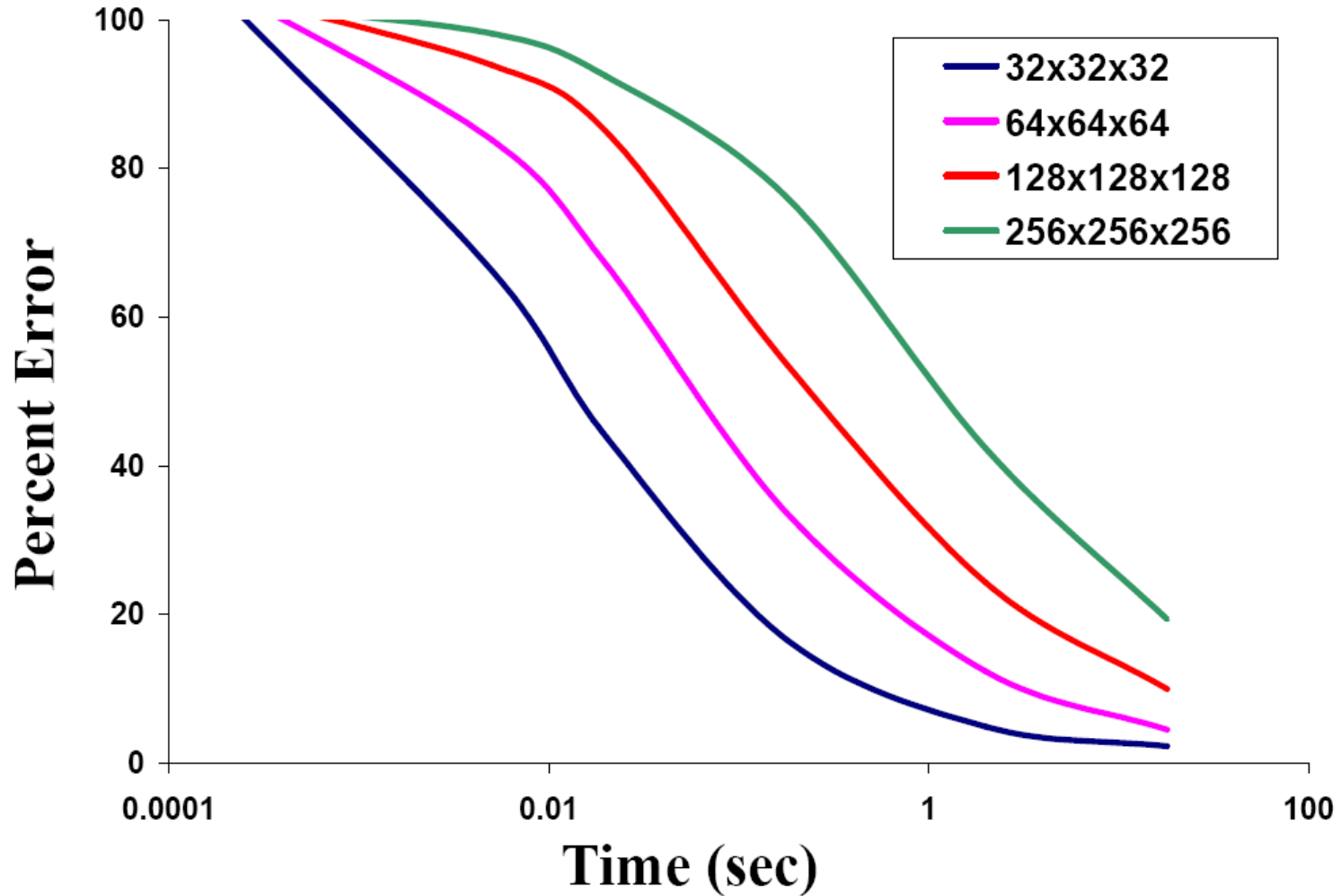
Input Model

Symmetry Transform

Symmetry Transform



Compute time:



Outline



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- Symmetry transform
- **Principal symmetries** ←

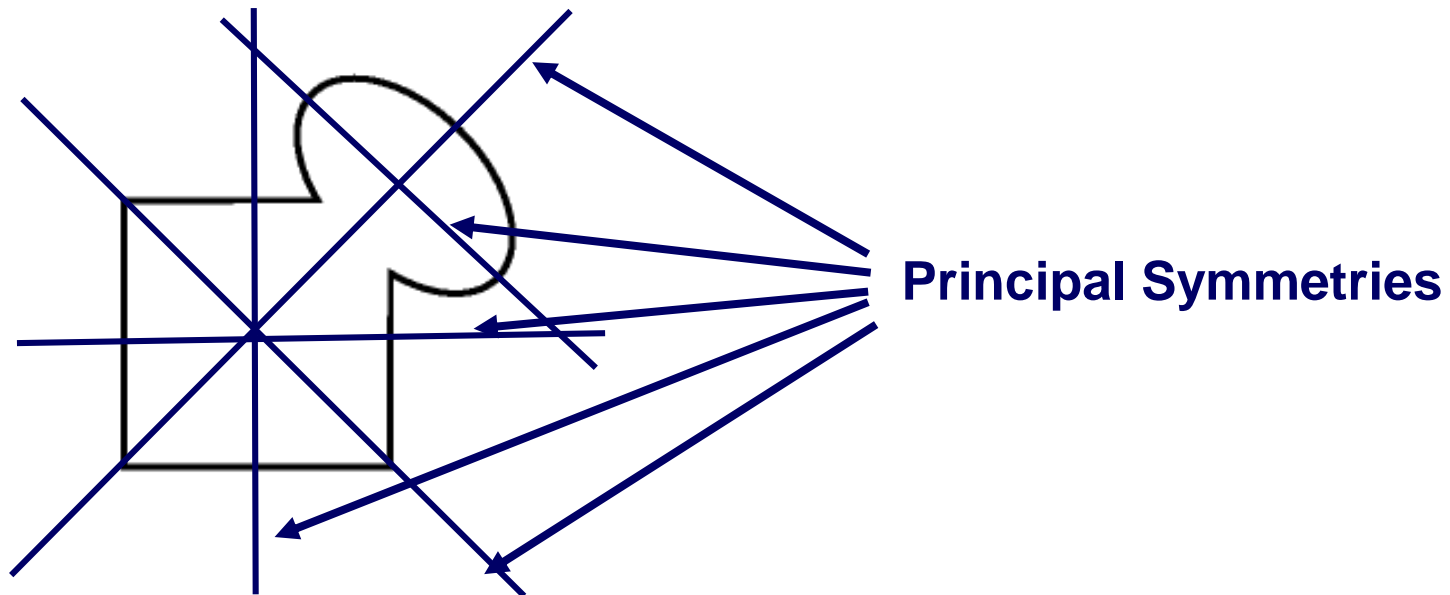
Applications

Conclusion



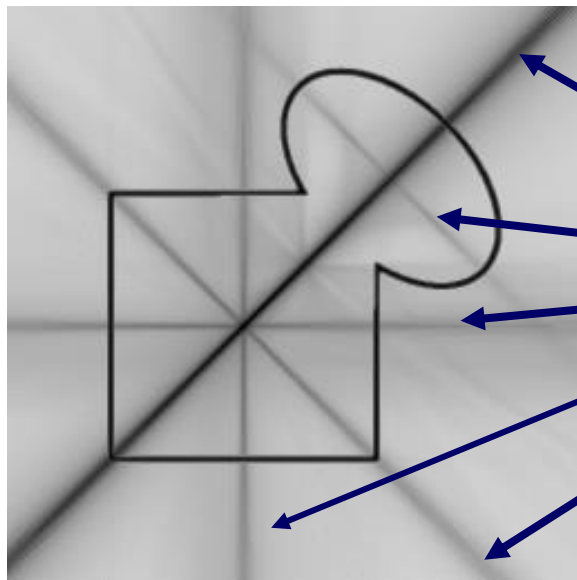
Principal Symmetries

Motivation: finding and representing significant symmetries is sufficient for many applications



Principal Symmetries

Observation: significant symmetries are usually local maxima of symmetry transformation

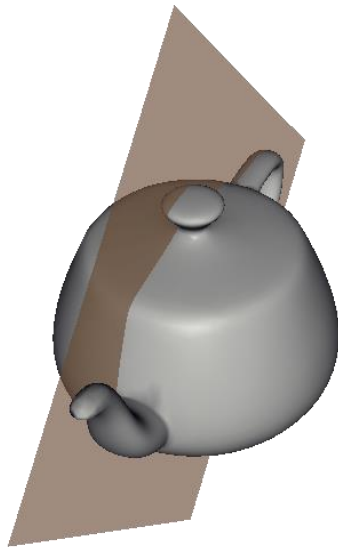


Principal Symmetries

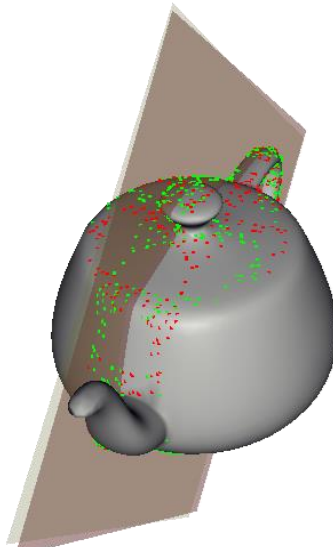
Principal Symmetries

Computing local maxima precisely:

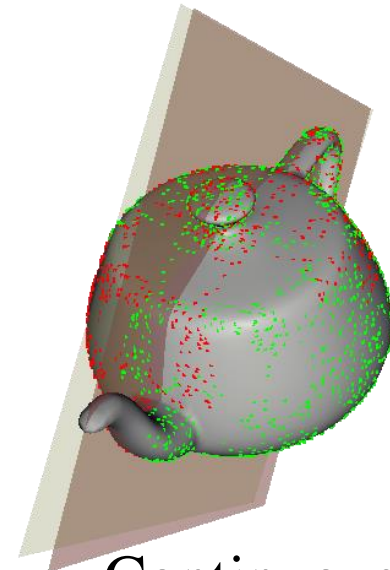
- Start from local maxima of discrete transform
- Establish closest point correspondences
- Refine transformation
- Iterate until find local maxima



Discrete
Local Maximum



Closest Point
Correspondences

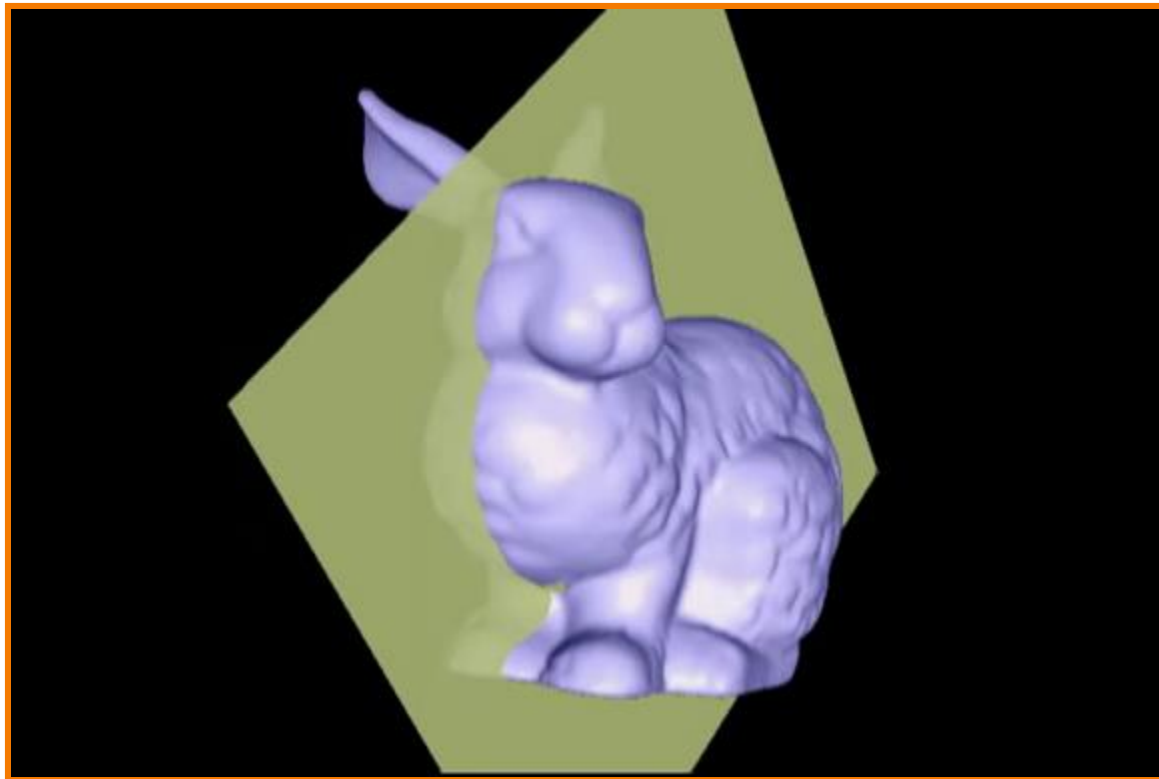


Continuous
Local Maximum

Principal Symmetries



Computing local maxima precisely:



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Applications ←

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Related Work

Exploiting symmetry in geometric processing

- Completion – Zabrodsky93, Thrun et al. 2005
- Alignment – Zabrodsky et al. 1995, Kazhdan et al. 2002
- Symmetrization - Zabrodsky et al. 1997, Mitra et al. 2007
- Feature detection – Reifeld et al. 1995
- Reverse engineering - Mills et al. 2001
- Instancing - Martinet et al. 2005
- Matching – Kazhdan et al. 2002, Gal et al. 2005
- Compression - Simari et al. 2006
- Segmentation - Mitra et al. 2006, Podolak et al. 2006
- Viewpoint selection – Podolak et al. 2006
- Editing – Mitra et al. 2006
- Simplification – Podolak et al. 2007

Applications



Alignment

Matching

Segmentation

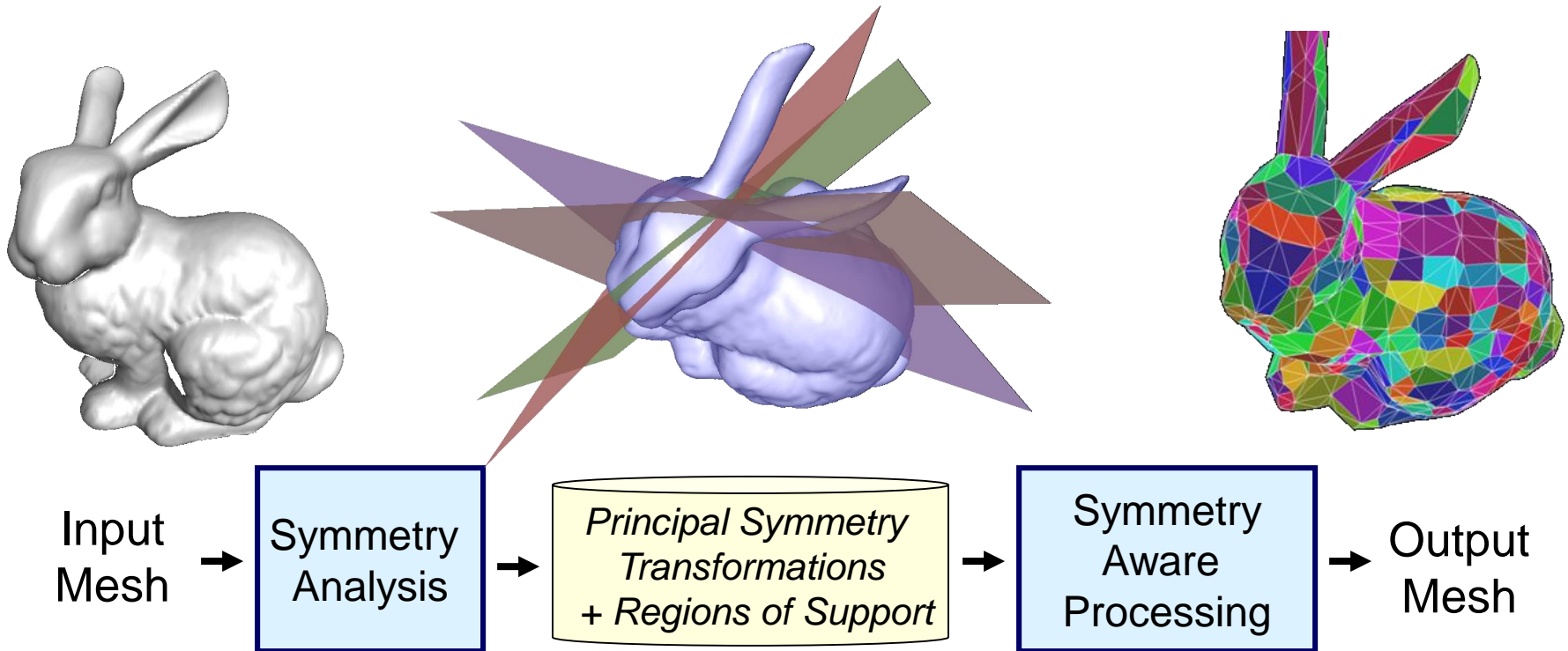
Viewpoint selection

Simplification

Beautification

Texture synthesis

Symmetry-Aware Processing



Applications



Alignment ←

Matching

Segmentation

Viewpoint selection

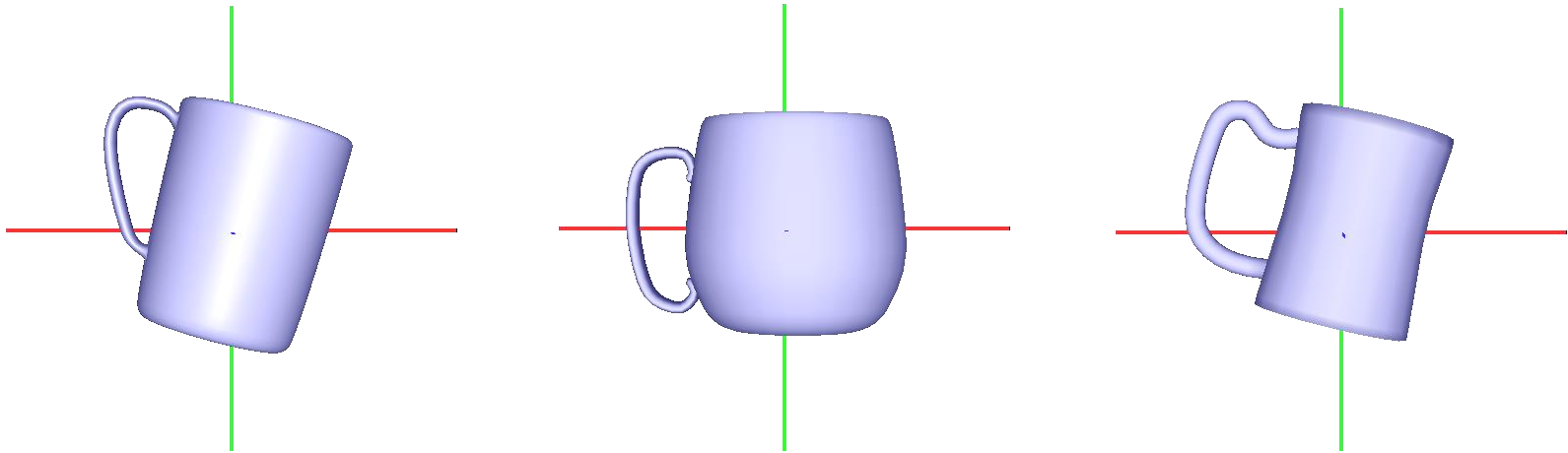
Simplification

Beautification

Texture synthesis

Application: Alignment

Motivation: registration, modeling, etc.

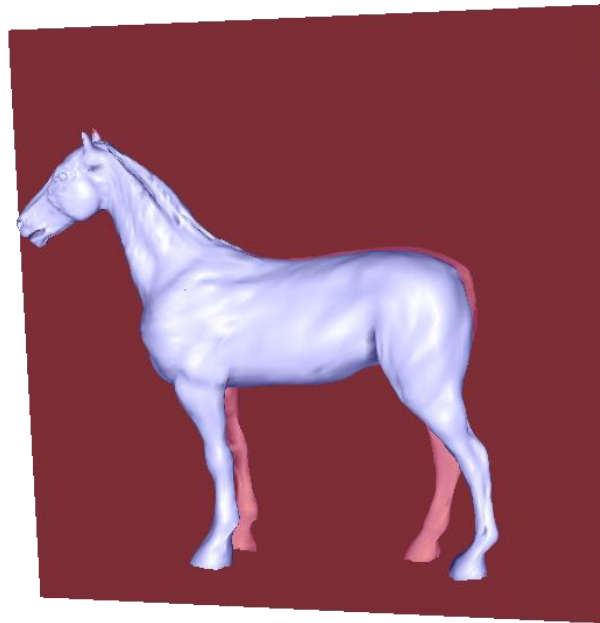


PCA Alignment

Application: Alignment



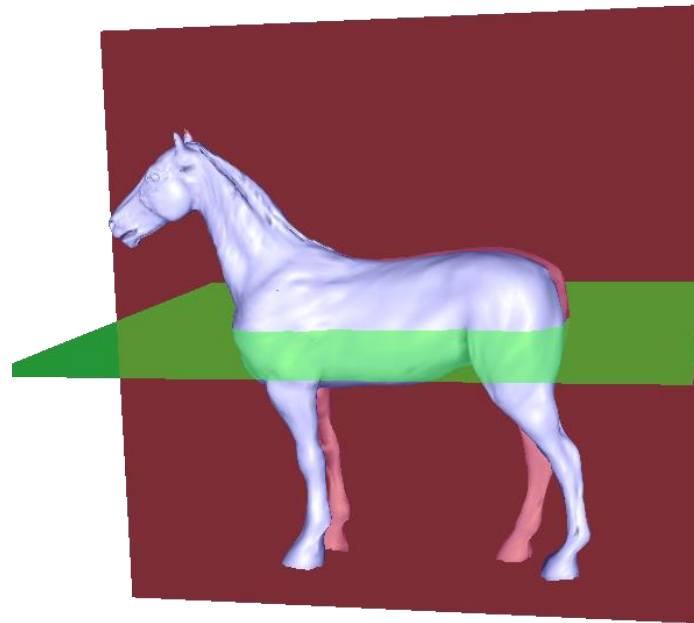
Approach: align planes with highest symmetries



Application: Alignment



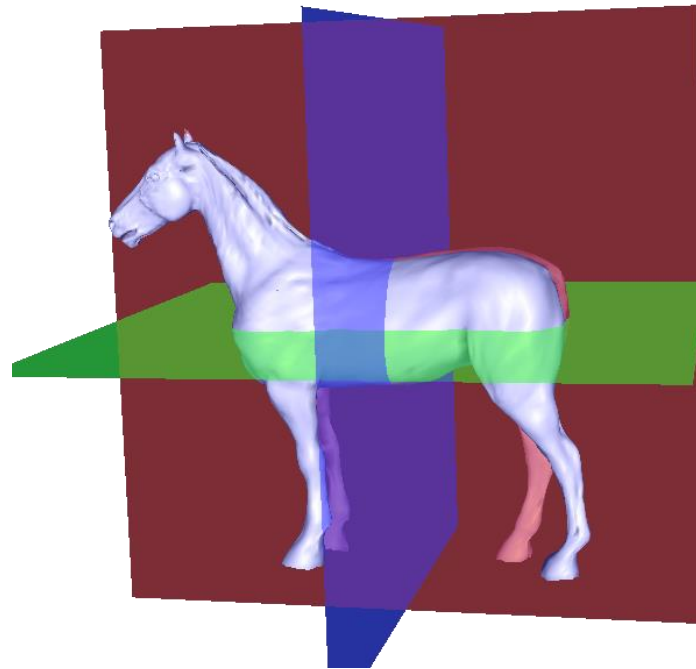
Approach: align planes with highest symmetries



Application: Alignment



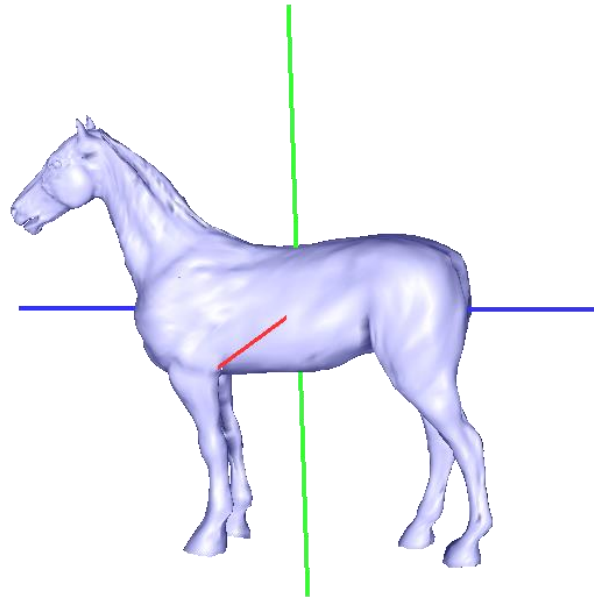
Approach: align planes with highest symmetries



Application: Alignment



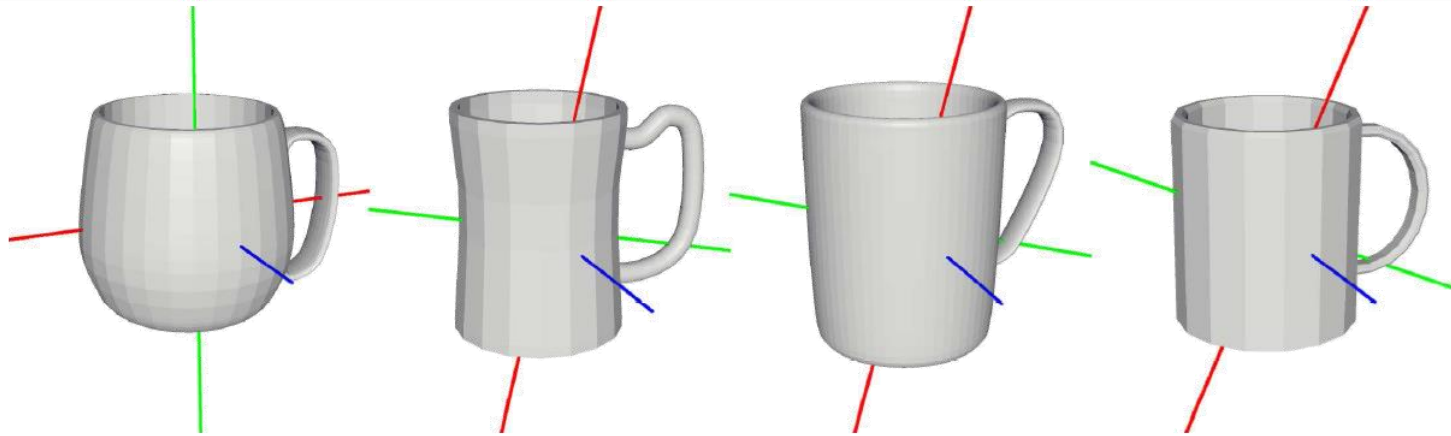
Approach: align planes with highest symmetries



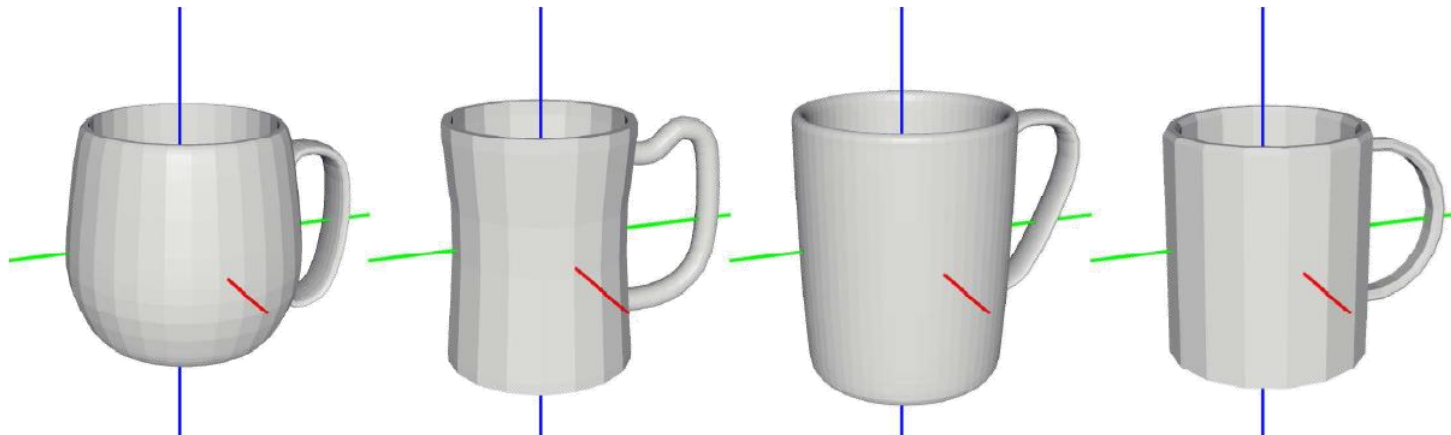
Application: Alignment



Results:



Center of mass and PCA

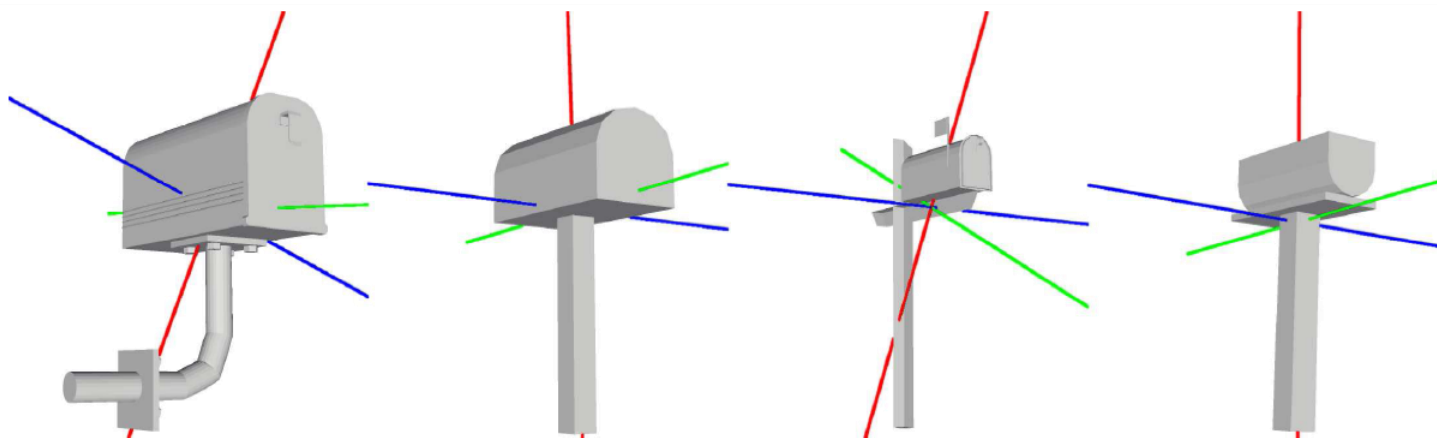


Center of symmetry and principal symmetry axes

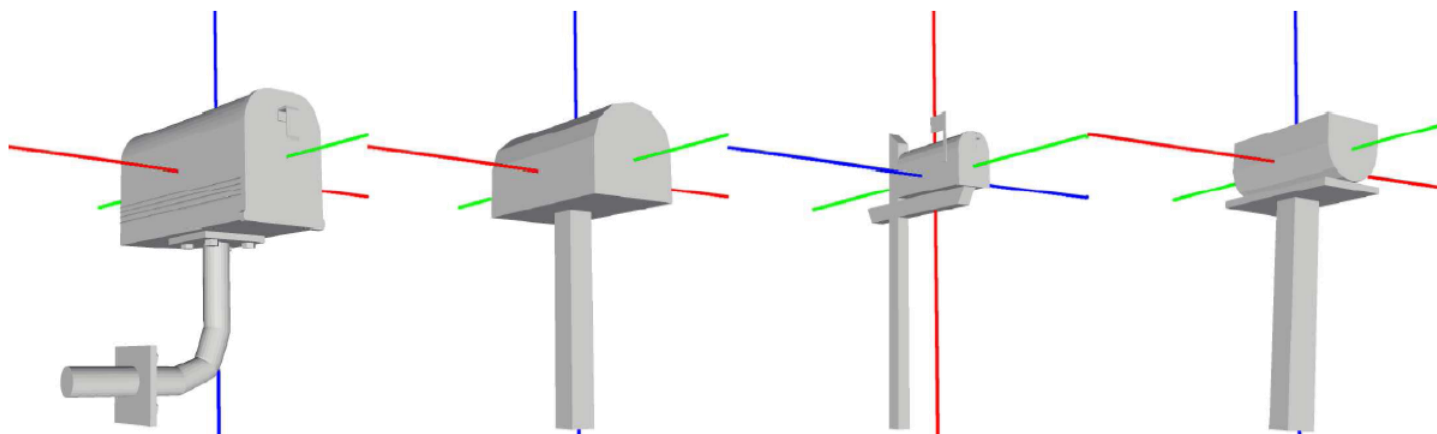
Application: Alignment



Results:



Center of mass and PCA

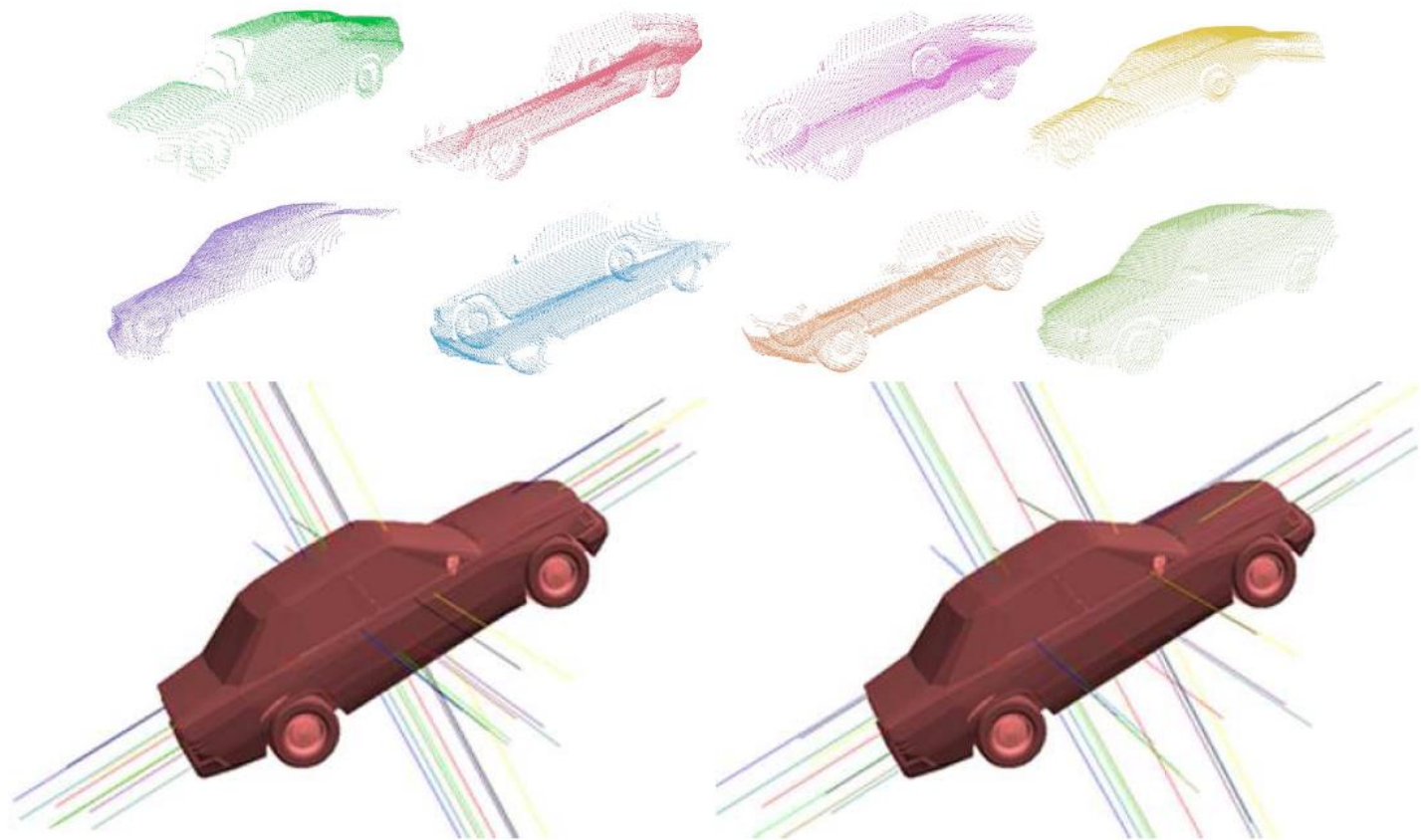


Center of symmetry and principal symmetry axes

Application: Alignment



Results:



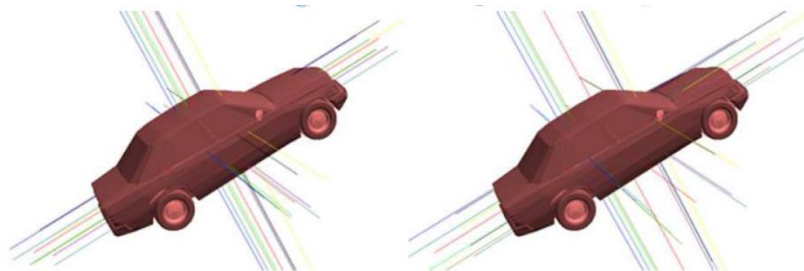
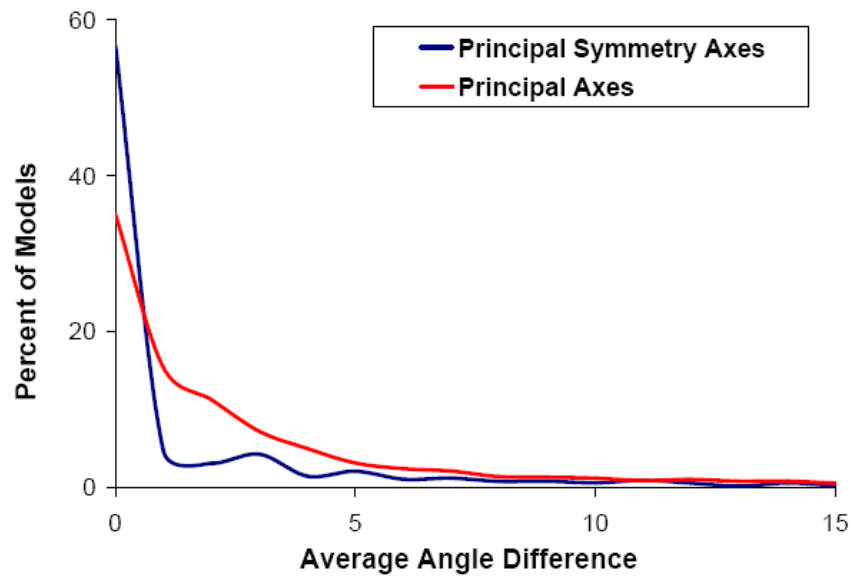
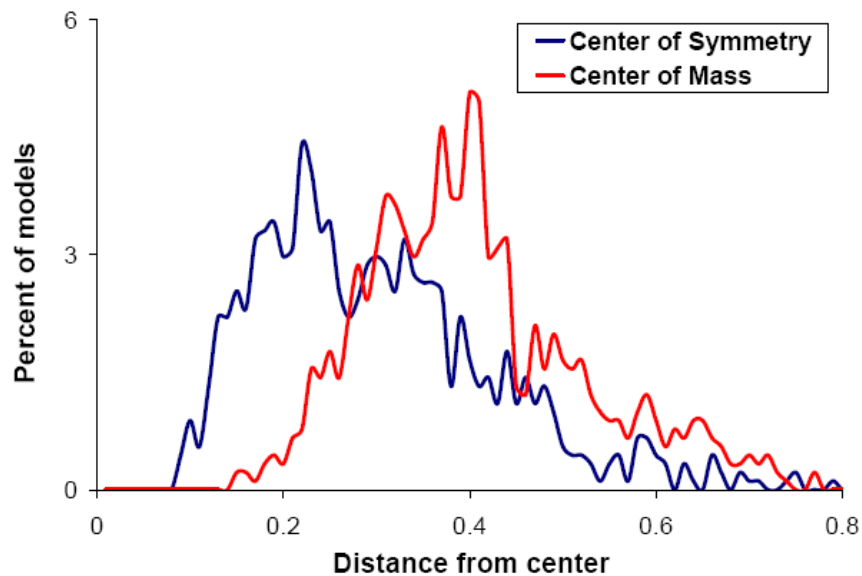
(Center of Symmetry)

(Center of Mass)

Application: Alignment



Results:



(Center of Symmetry)

(Center of Mass)

Applications



Alignment

Matching ←

Segmentation

Viewpoint selection

Simplification

Beautification

Texture synthesis

Application: Matching



Motivation: similarity search of database



Query



Database



Best Match



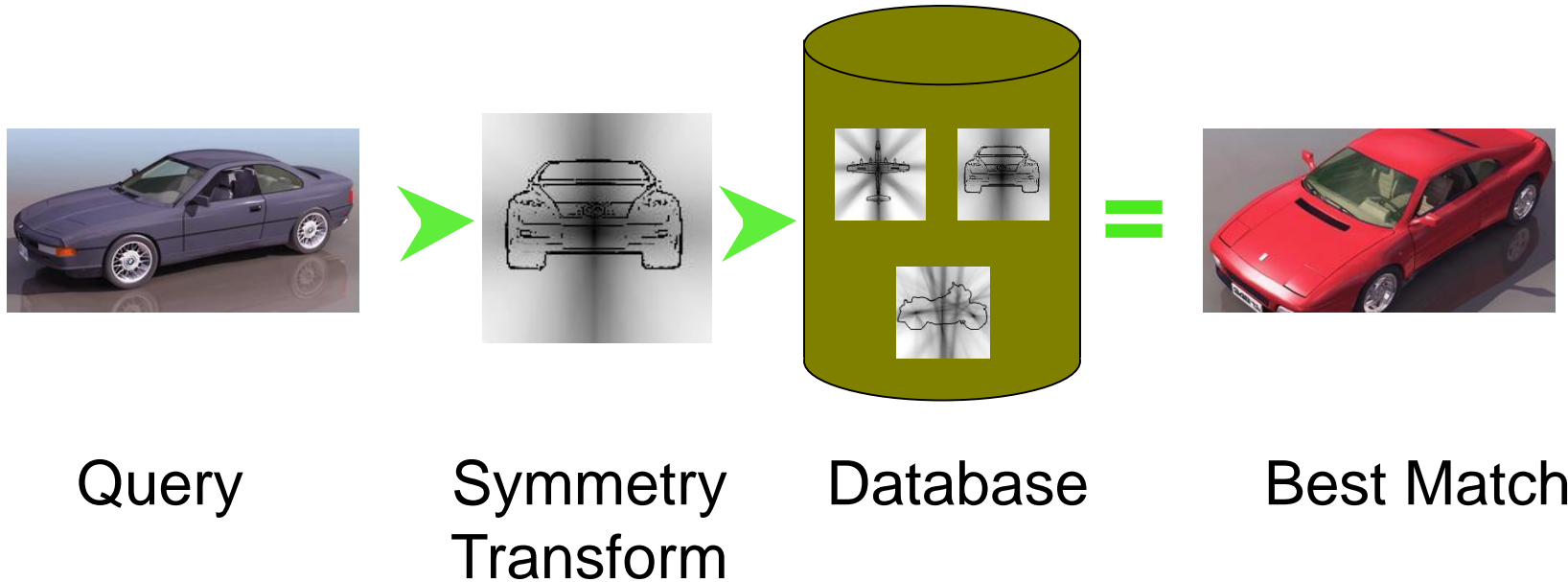
Application: Matching

Observation: symmetry is more consistent than shape for some object classes



Application: Matching

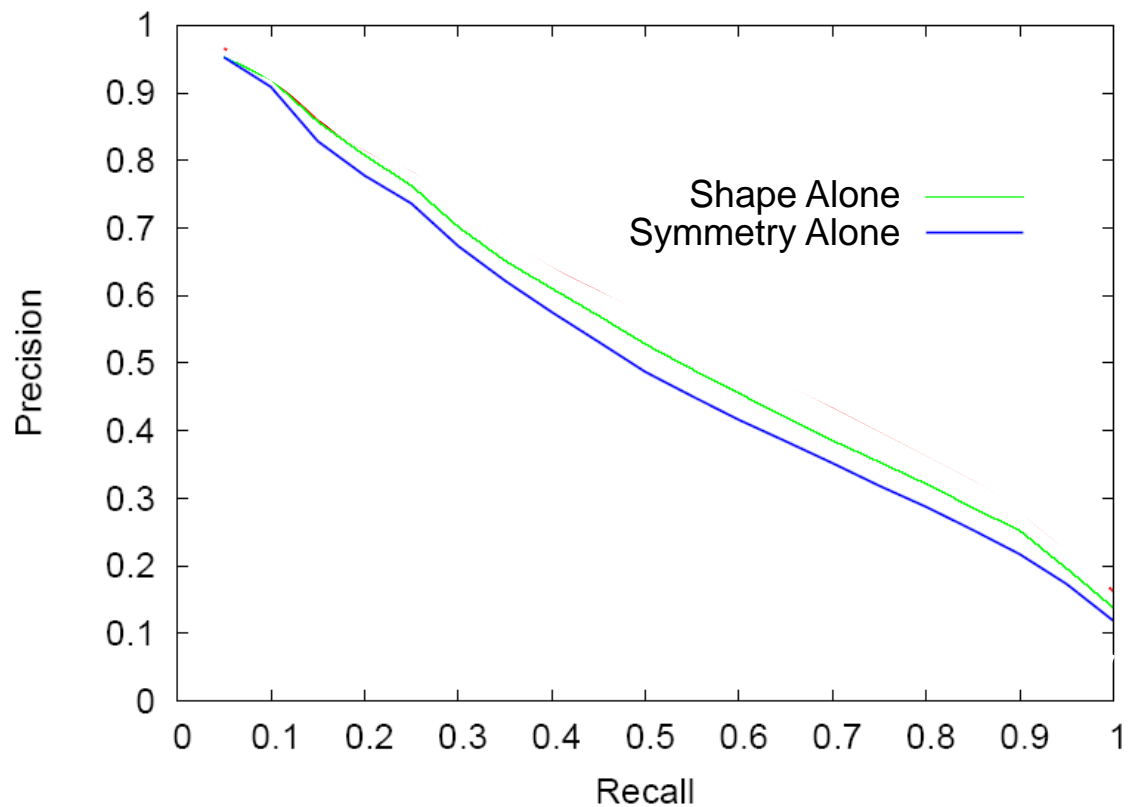
Approach: use symmetry transform (or descriptor) as shape descriptor



Application: Matching



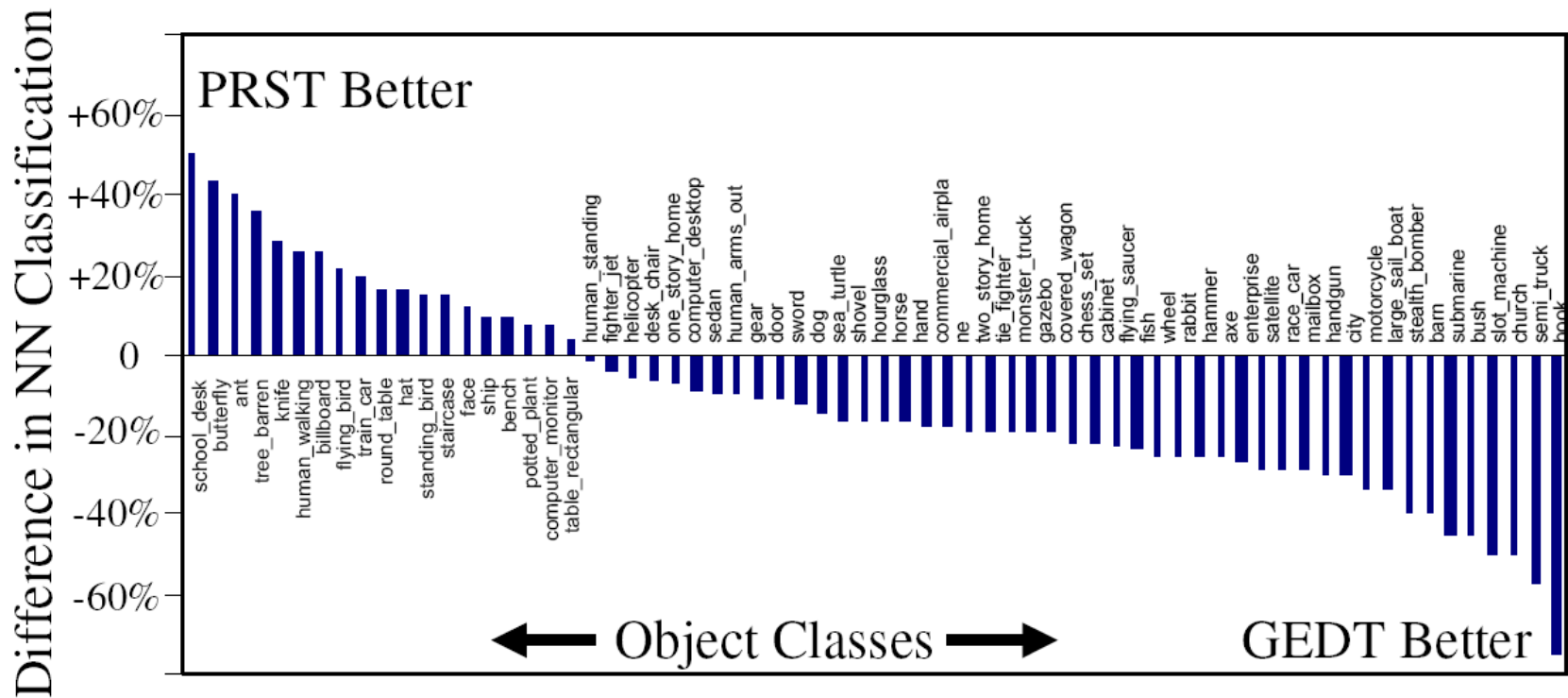
Results: symmetry is not as discriminating as shape



Application: Matching



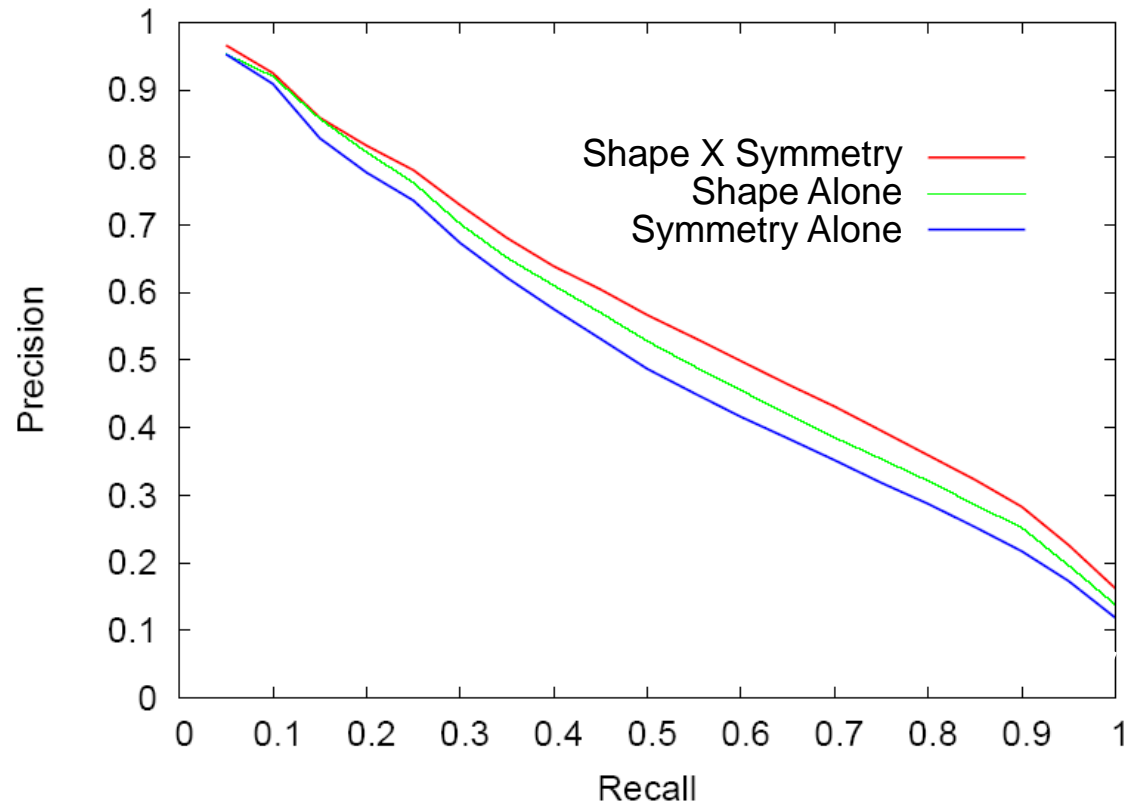
Results: symmetry is not as discriminating as shape, but it is better for some classes



Application: Matching



Results: symmetry is not as discriminating as shape, but it is better for some classes, and so the two together are better than either alone





Applications

Alignment

Matching

Segmentation ←

Viewpoint selection

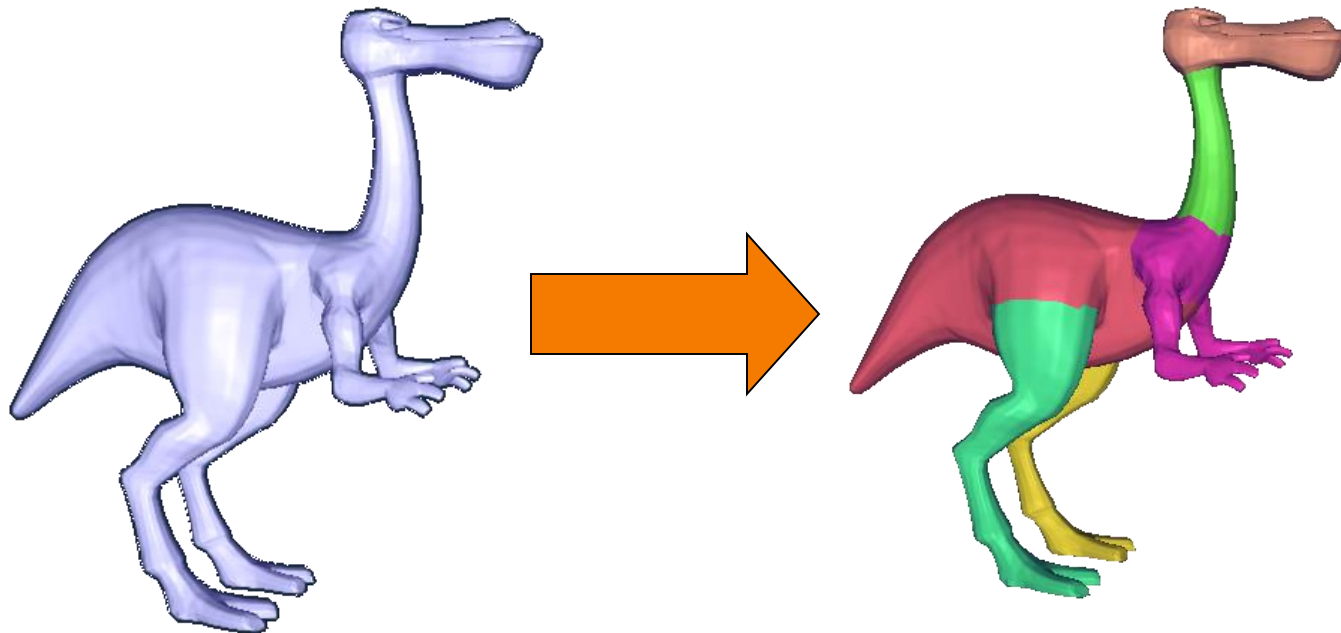
Simplification

Beautification

Texture synthesis

Application: Segmentation

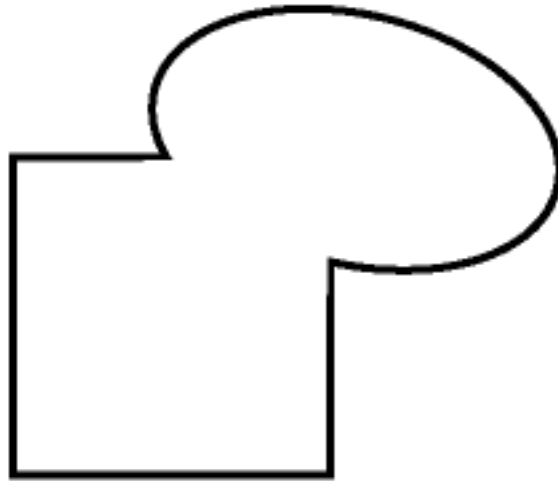
Motivation: animation, modeling by parts, etc.



Application: Segmentation



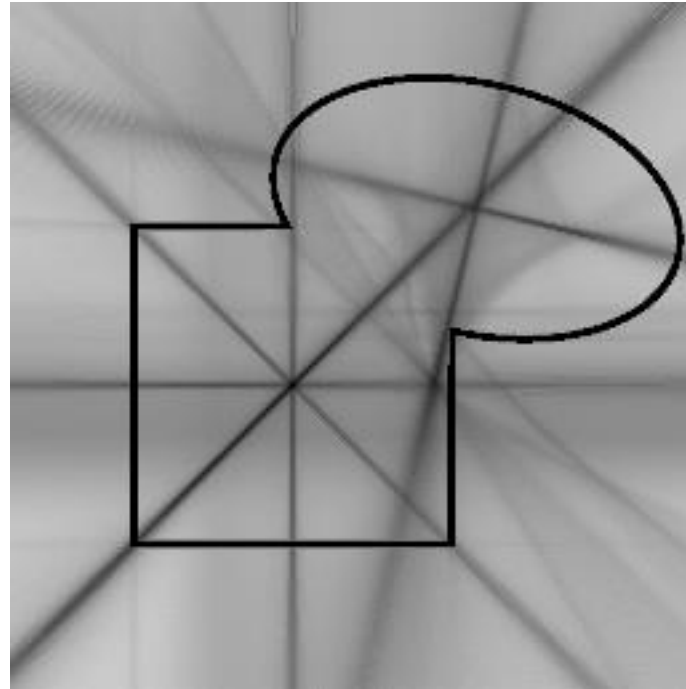
Observation: distinct parts have strong local symmetries not shared by other parts



Application: Segmentation



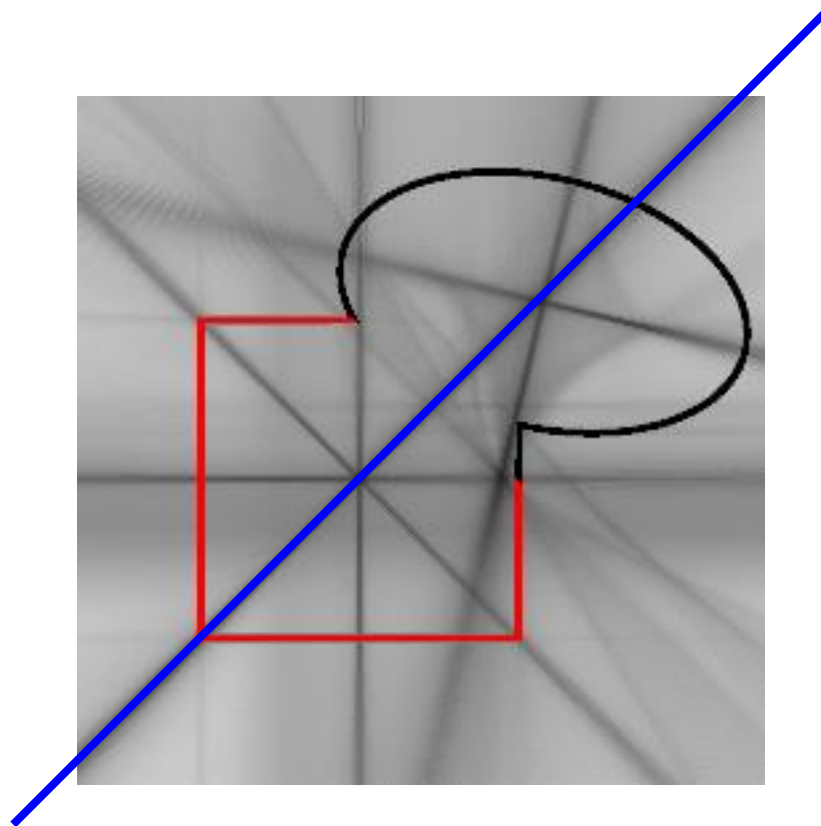
Observation: distinct parts have strong local symmetries not shared by other parts





Application: Segmentation

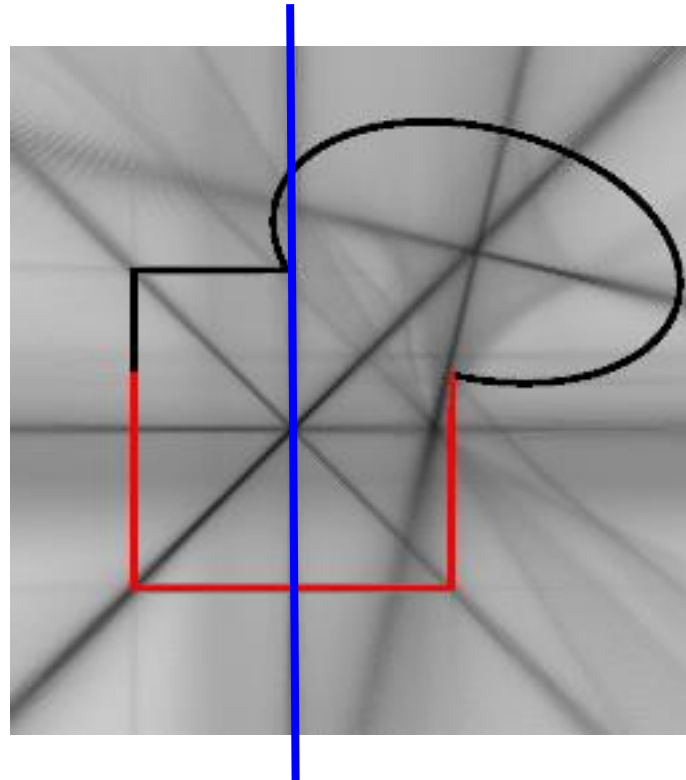
Observation: distinct parts have strong local symmetries not shared by other parts





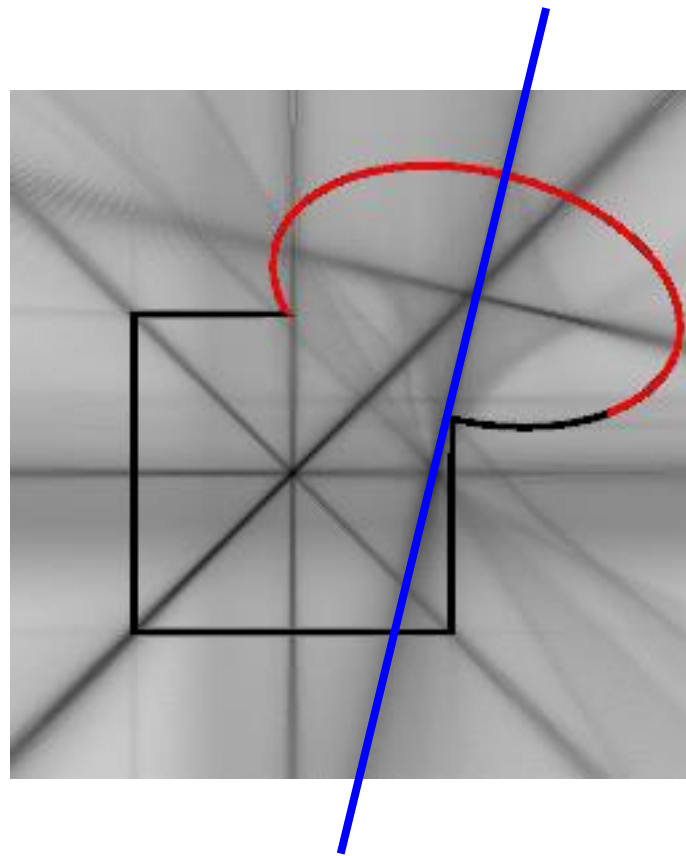
Application: Segmentation

Observation: distinct parts have strong local symmetries not shared by other parts



Application: Segmentation

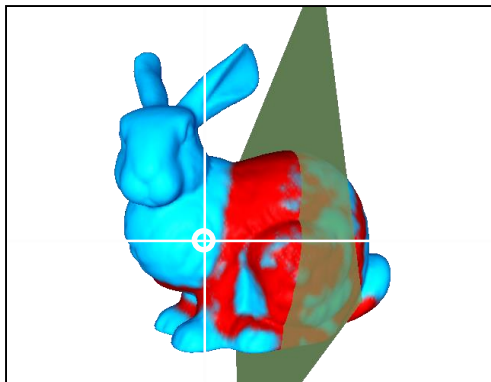
Observation: distinct parts have strong local symmetries not shared by other parts



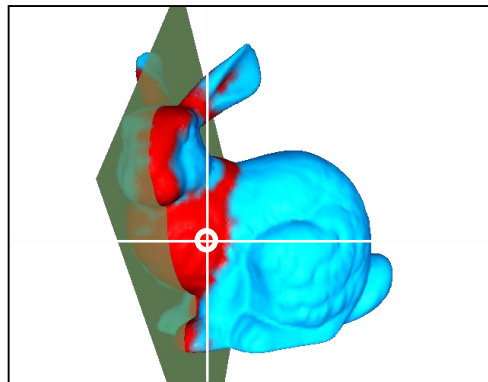
Application: Segmentation



Approach: cluster points on the surface by how much they support different symmetries

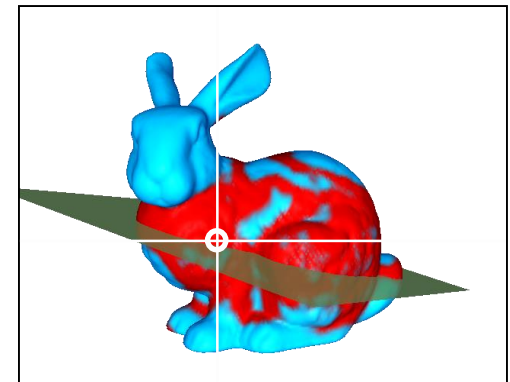


Support = 0.1



Support = 0.5

.....

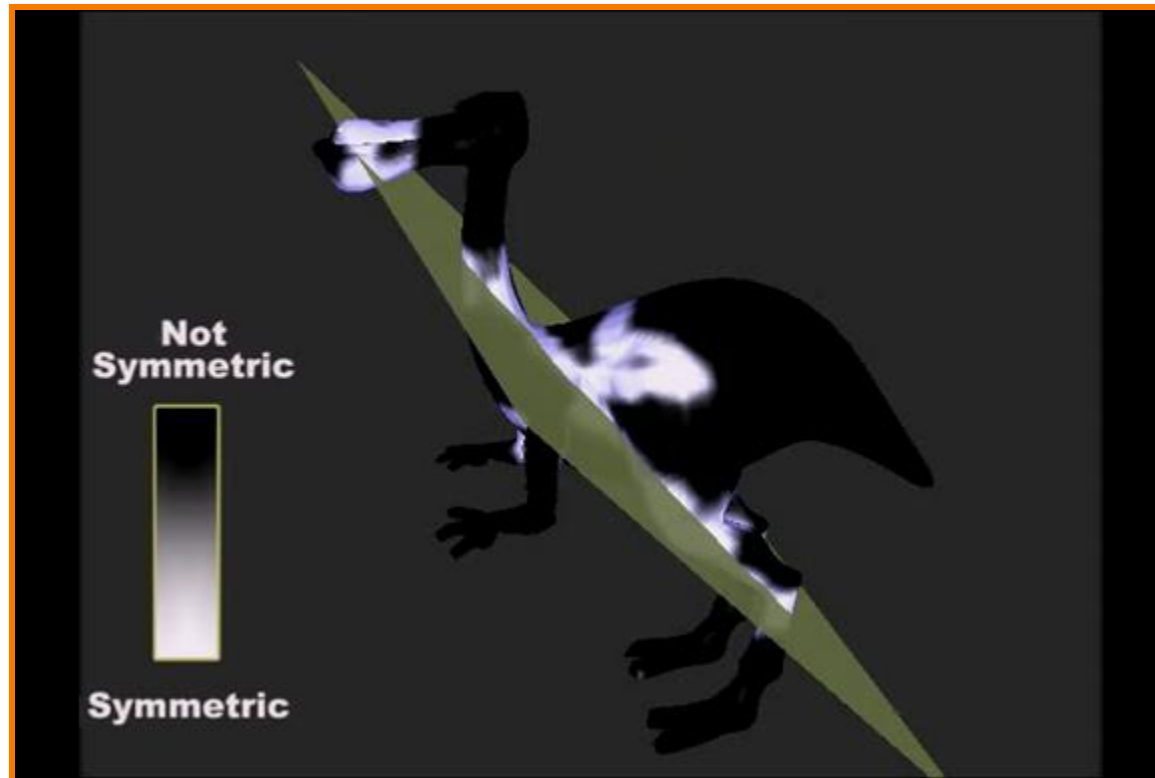


Support = 0.9

Symmetry Vector = { 0.1 , 0.5 , , 0.9 }

Application: Segmentation

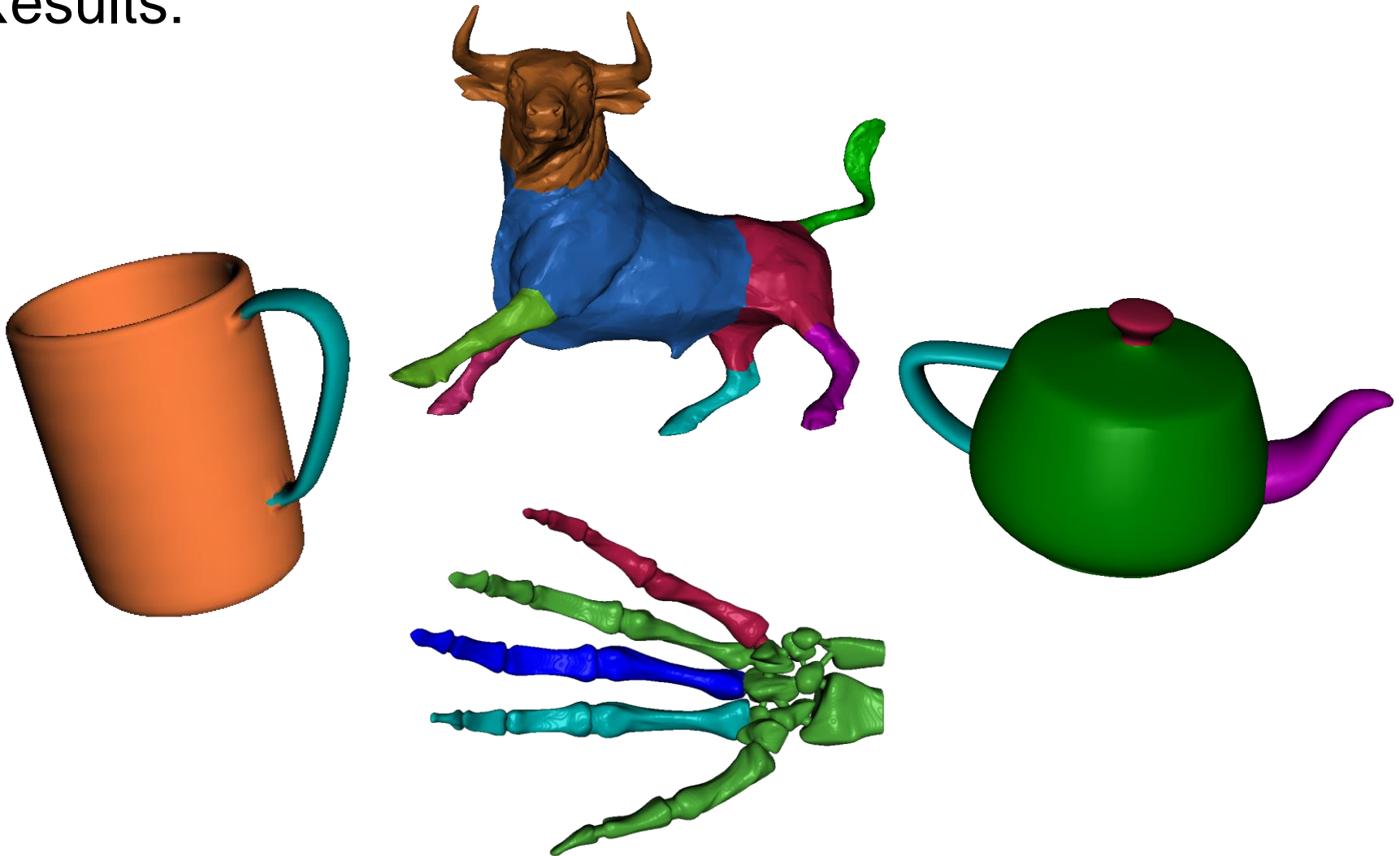
Approach: cluster points on the surface by how much they support different symmetries



Application: Segmentation



Results:





Applications

Alignment

Matching

Segmentation

Viewpoint selection ←

Simplification

Beautification

Texture synthesis

Application: Viewpoint Selection



Motivation: visualization, icon generation, etc.



Application: Viewpoint Selection



Observation: symmetry represents redundancy



Bad Viewpoint

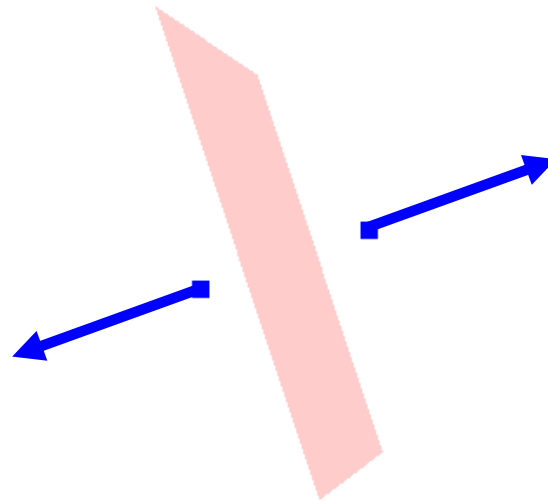
Application: Viewpoint Selection



Approach: Minimize visible symmetry

- Every plane of symmetry votes for a viewing direction perpendicular to it

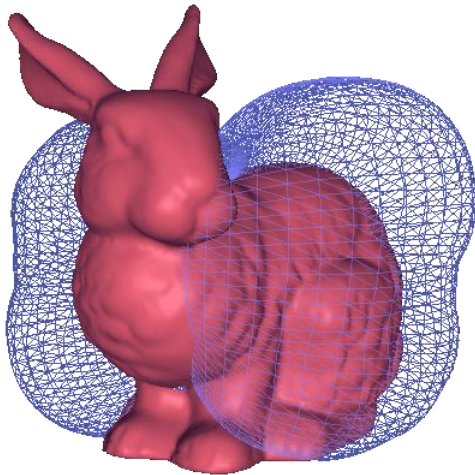
**Best Viewing
Directions**



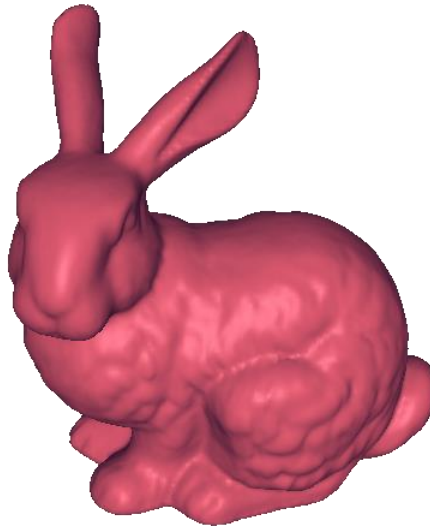
Application: Viewpoint Selection



Results:



Viewpoint
Function



Best
Viewpoint

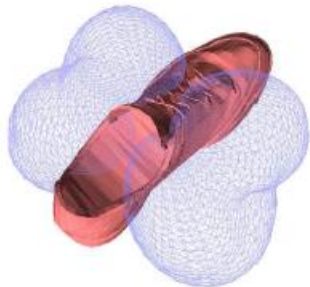
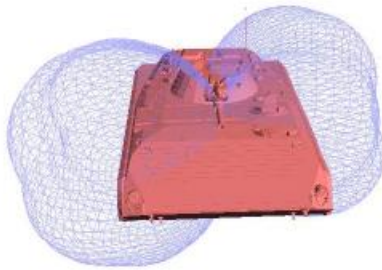
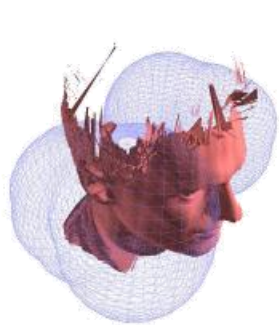


Worst
Viewpoint

Application: Viewpoint Selection



Results:



Viewpoint Function

Best Viewpoint

Worst Viewpoint

Applications



Alignment

Matching

Segmentation

Viewpoint selection

Simplification ←

Beautification

Texture synthesis

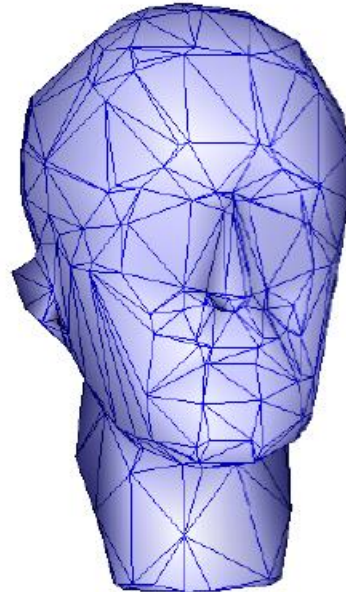
Application: Simplification



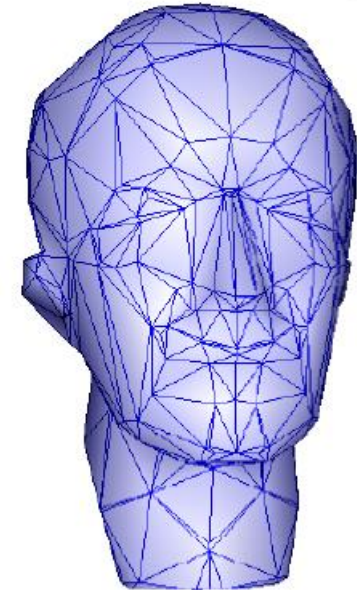
Motivation: preserve symmetry when simplify mesh



Input
Mesh



Standard
Simplification



Symmetric
Simplification

Application: Simplification



Approach 1: detect (approximate) symmetries and then preserve them as decimate



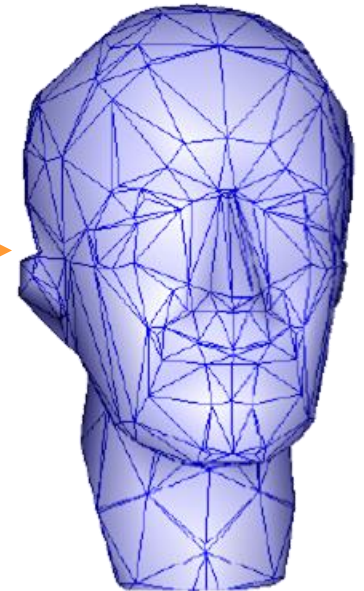
Input Mesh

Symmetry Analysis



Symmetric Correspondences

Symmetry Aware Decimation

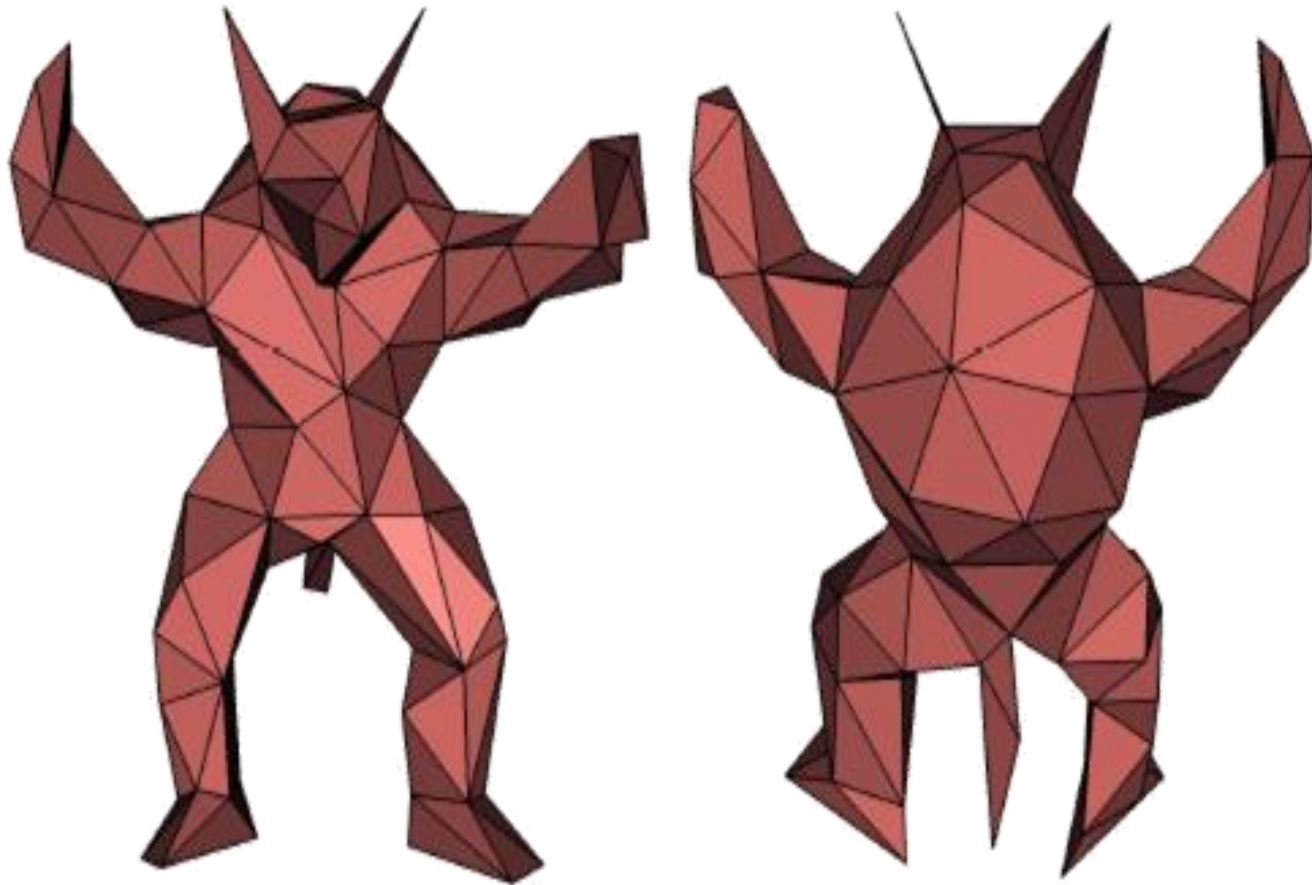


Symmetric Simplification

Application: Simplification



Results:

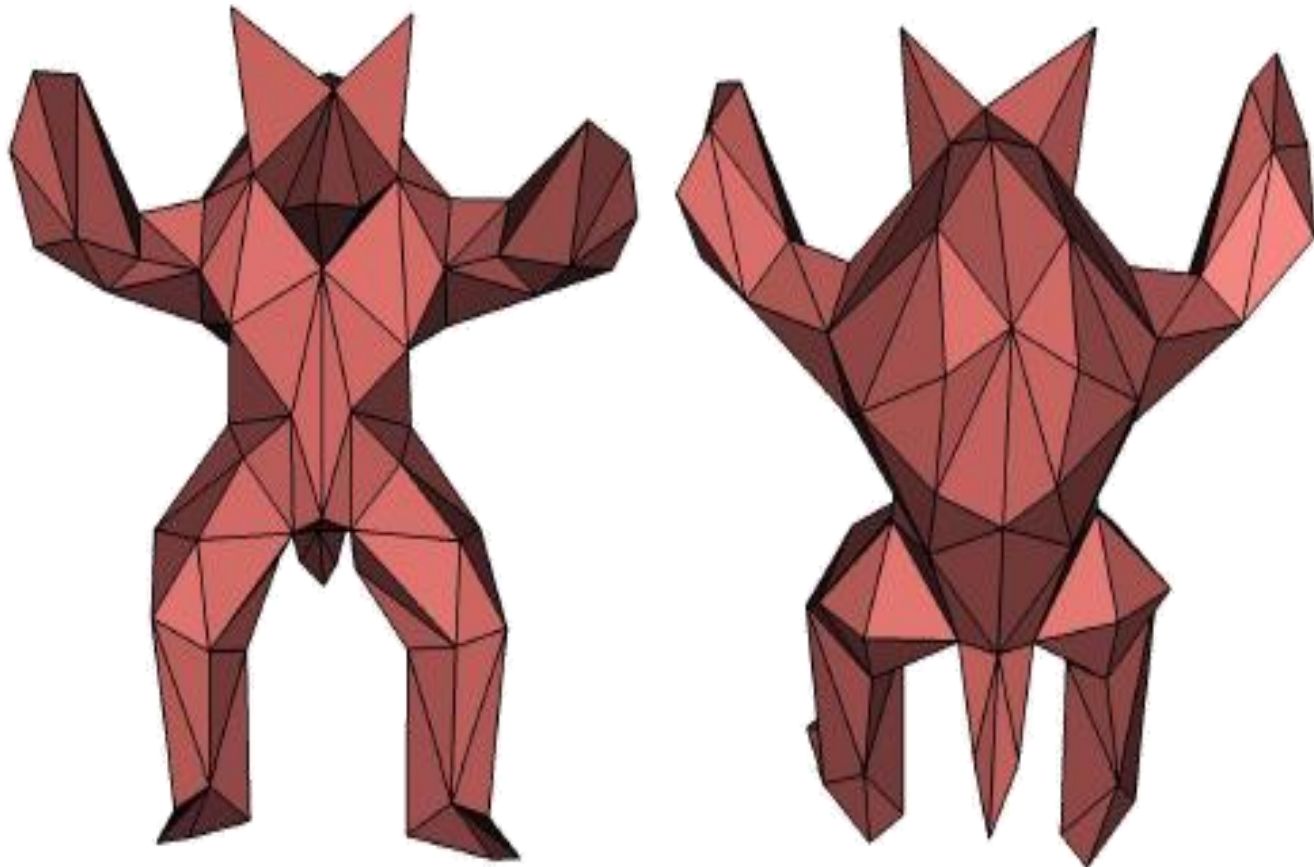


Original Qslim

Application: Simplification



Results:

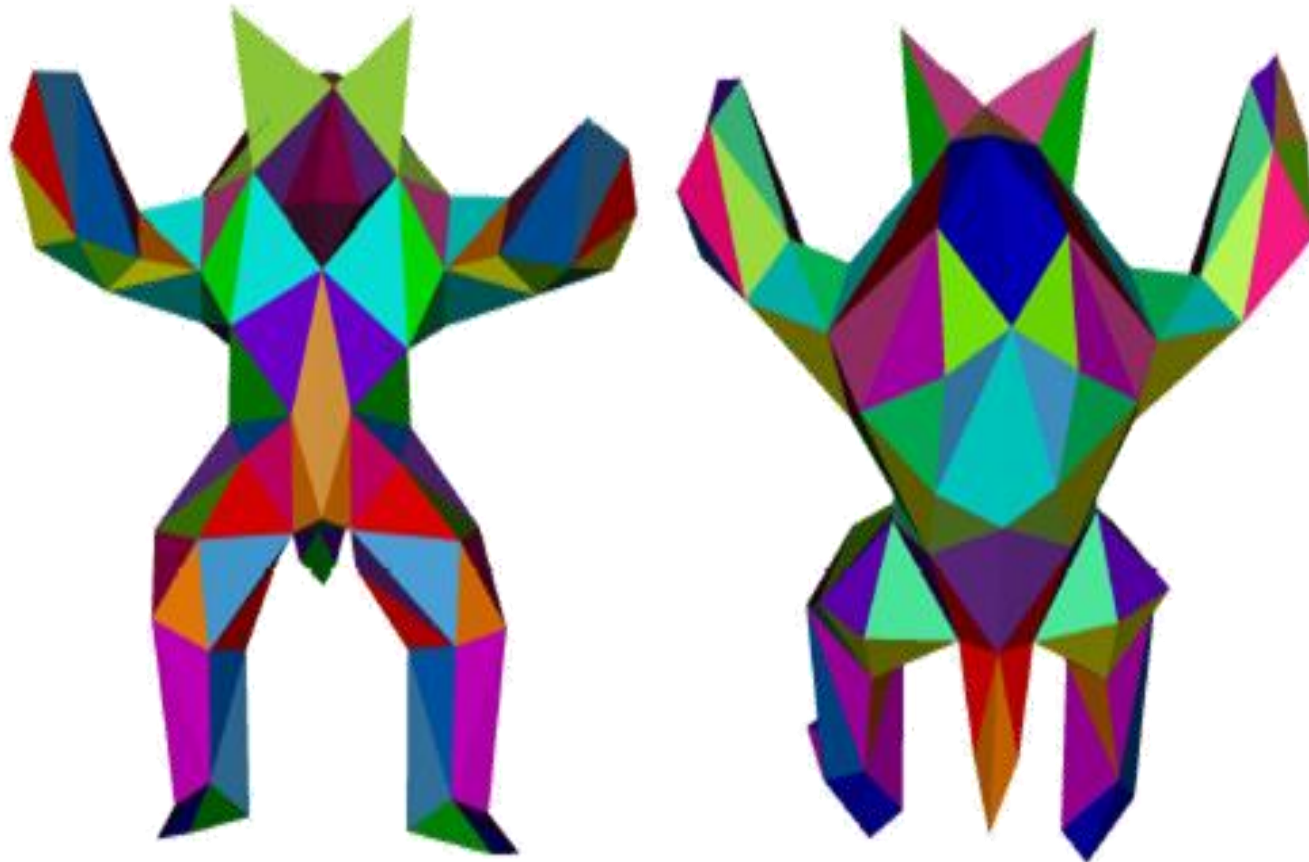


Symmetry-Preserving Qslim

Application: Simplification



Results:



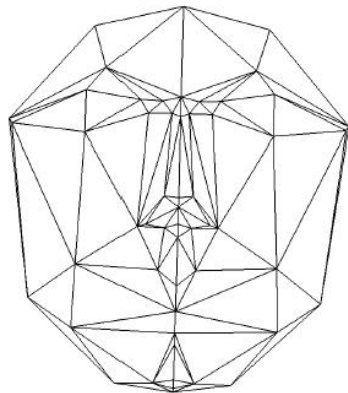
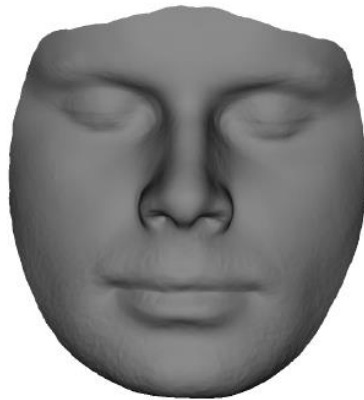
Symmetry-Preserving Qslim

Application: Parameterization

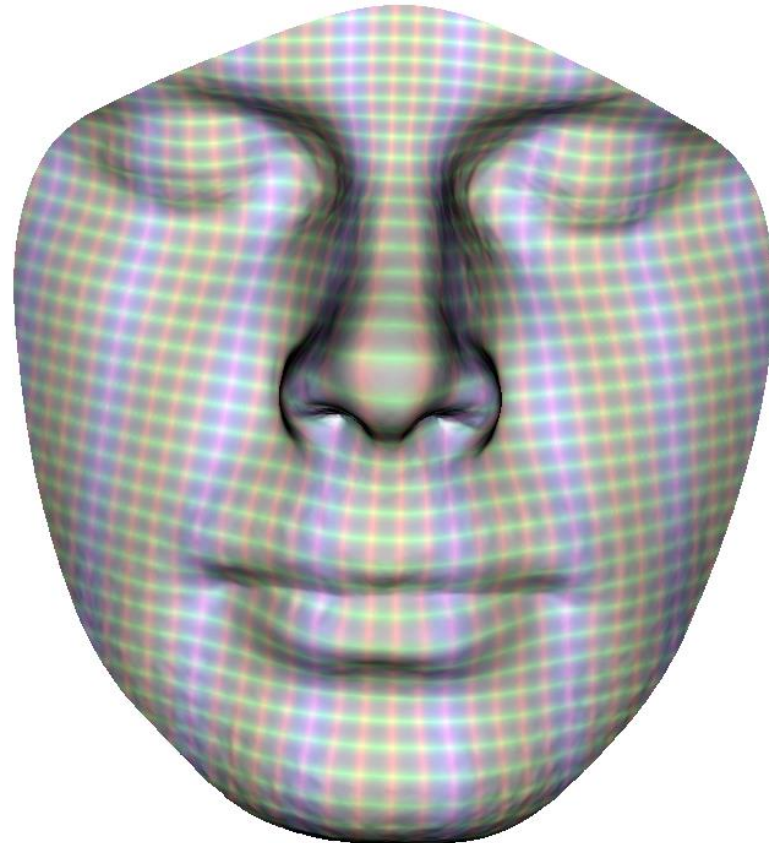


Preserve symmetries in base domain

Input Mesh



Base Domain



Parameterization

Applications



Alignment

Matching

Segmentation

Viewpoint selection

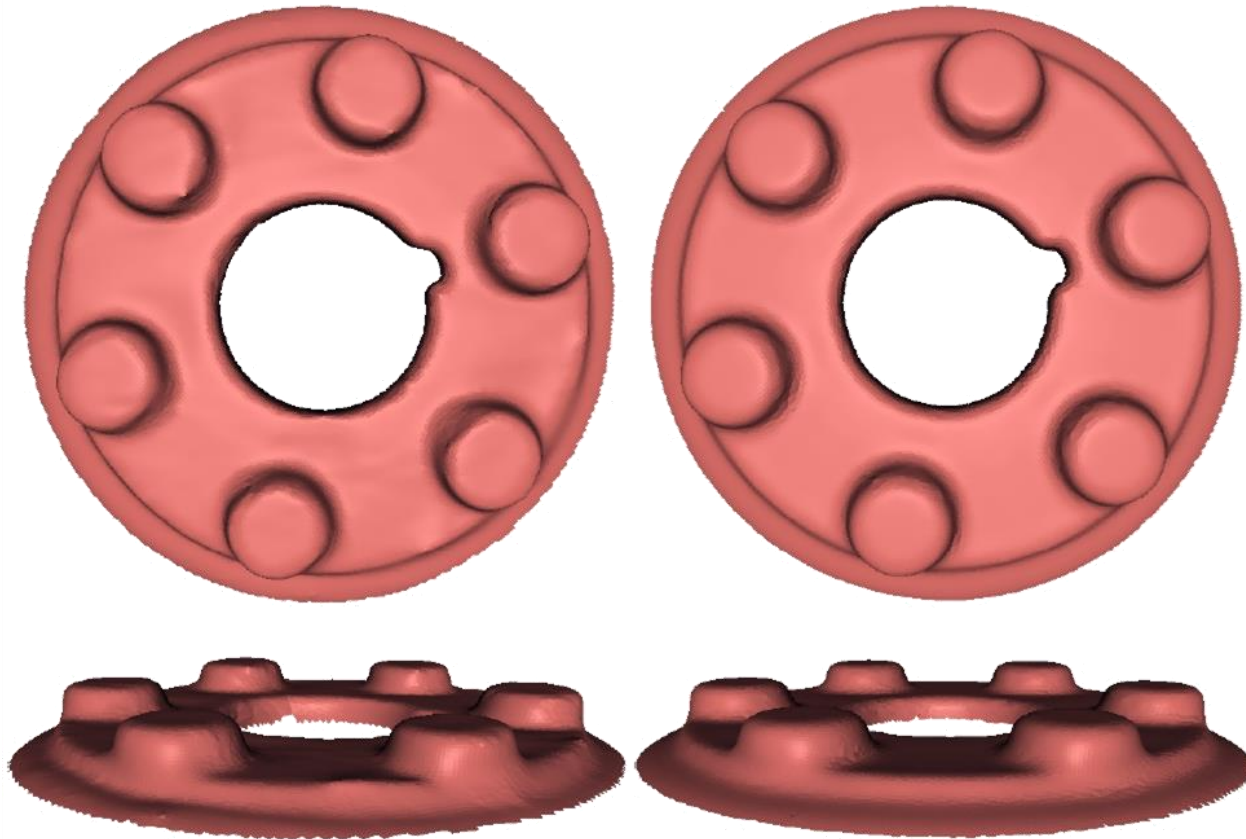
Simplification

Beautification ←

Texture synthesis

Application: Beautification

Goal: make meshes of symmetric objects perfectly symmetric



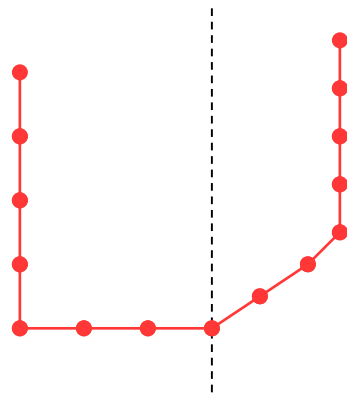
Input Mesh

Symmetric Mesh

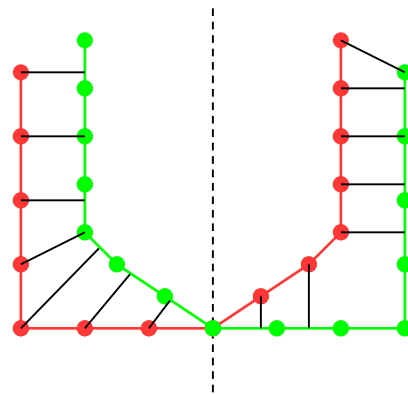
Application: Beautification



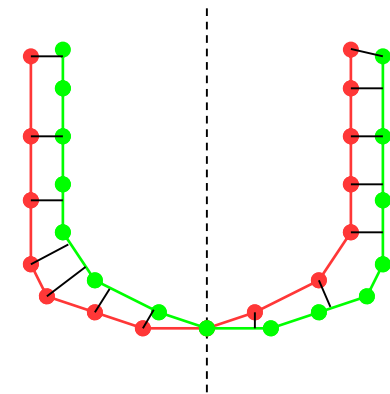
Approach: iterative non-rigid deformation to align symmetric points (symmetrization)



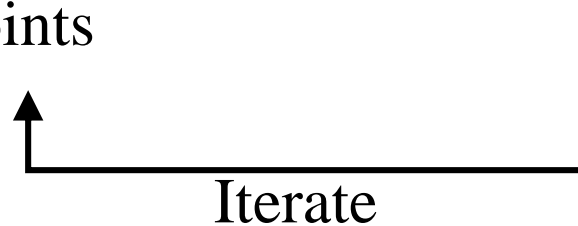
Input shape



Correspondences
Between
Symmetric
Points

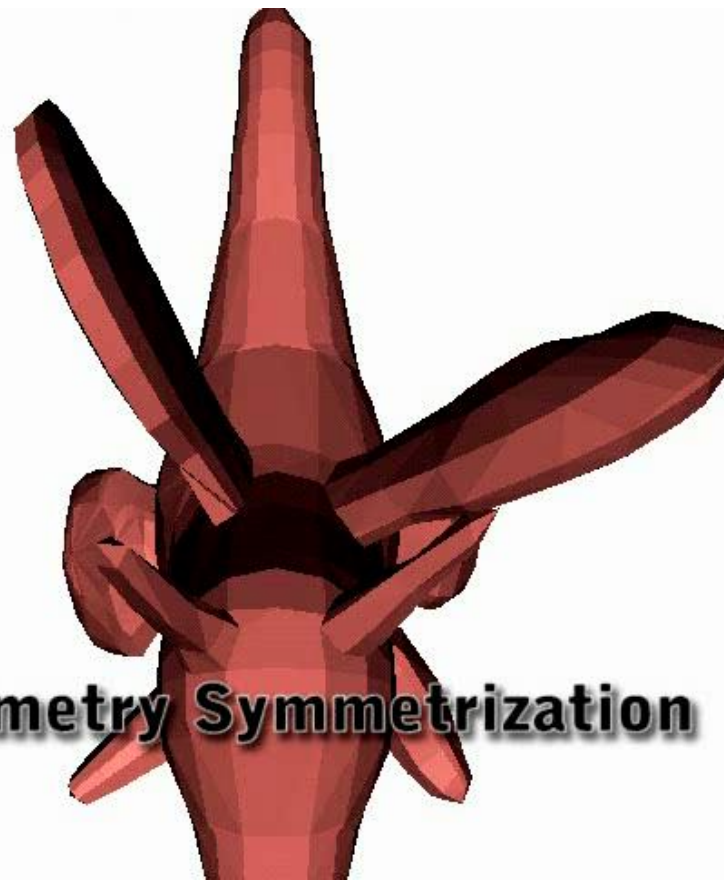


Shape Preserving
Deformation to
Enhance Symmetry



Application: Beautification

Approach: iterative non-rigid deformation to align symmetric points (symmetrization)

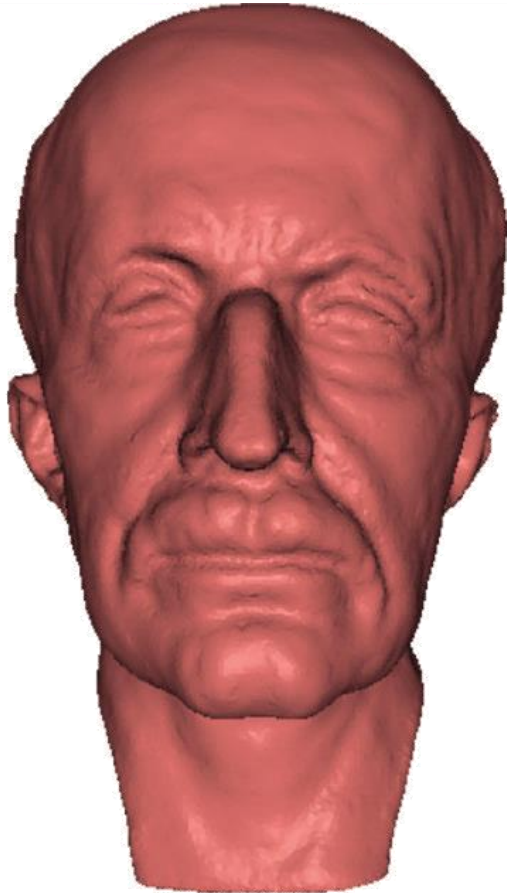


Geometry Symmetrization

Application: Beautification



Results:



Input Mesh



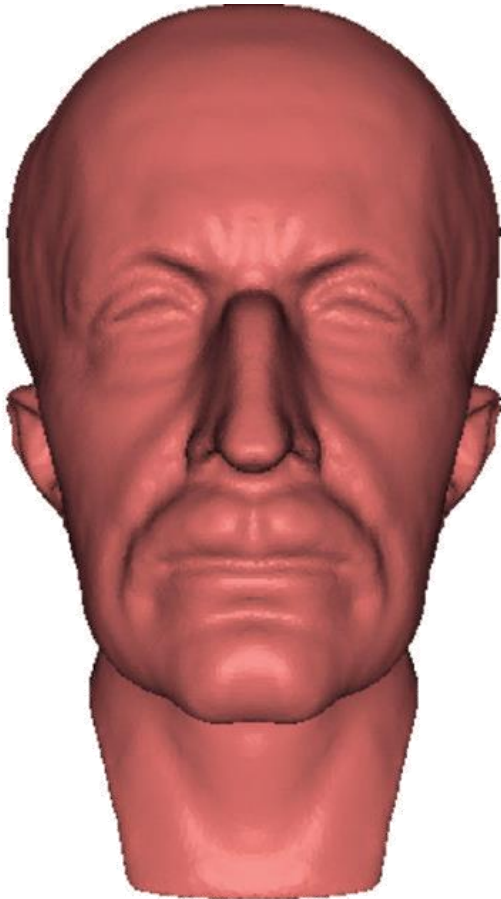
Mesh overlaid with its reflection



Application: Beautification



Results:



Symmetrized Mesh



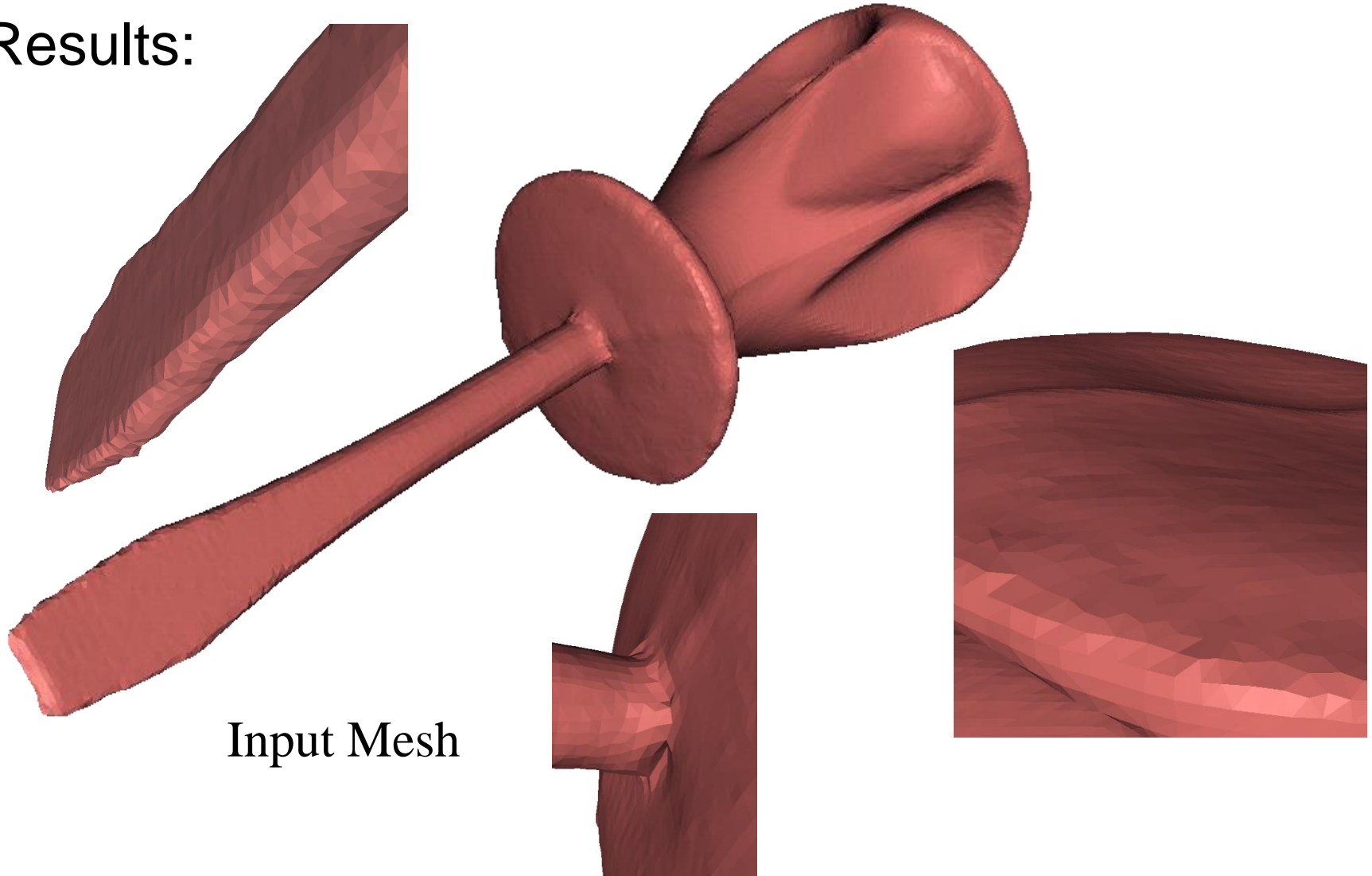
Mesh overlaid with its reflection



Application: Beautification



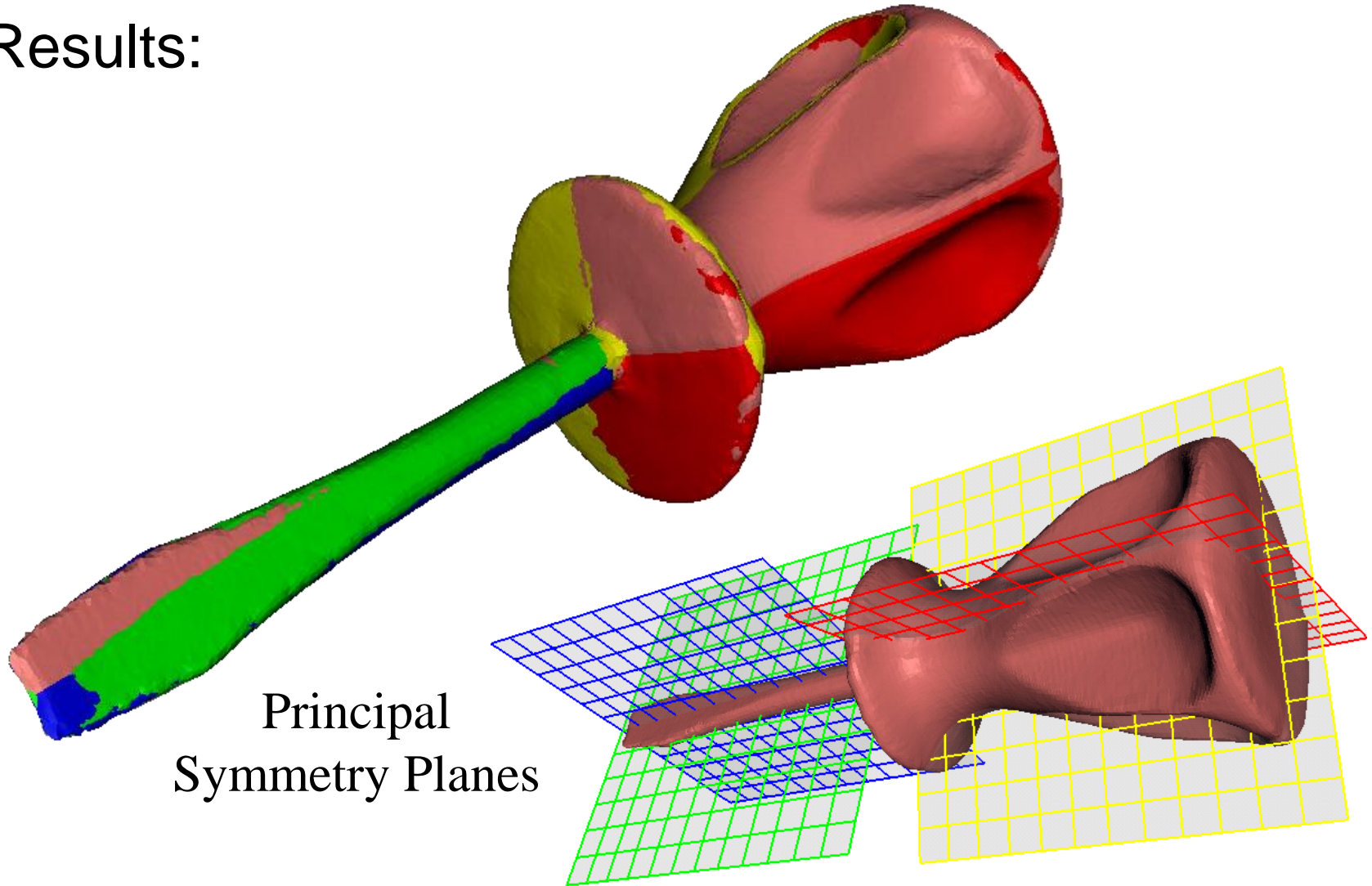
Results:



Application: Beautification



Results:

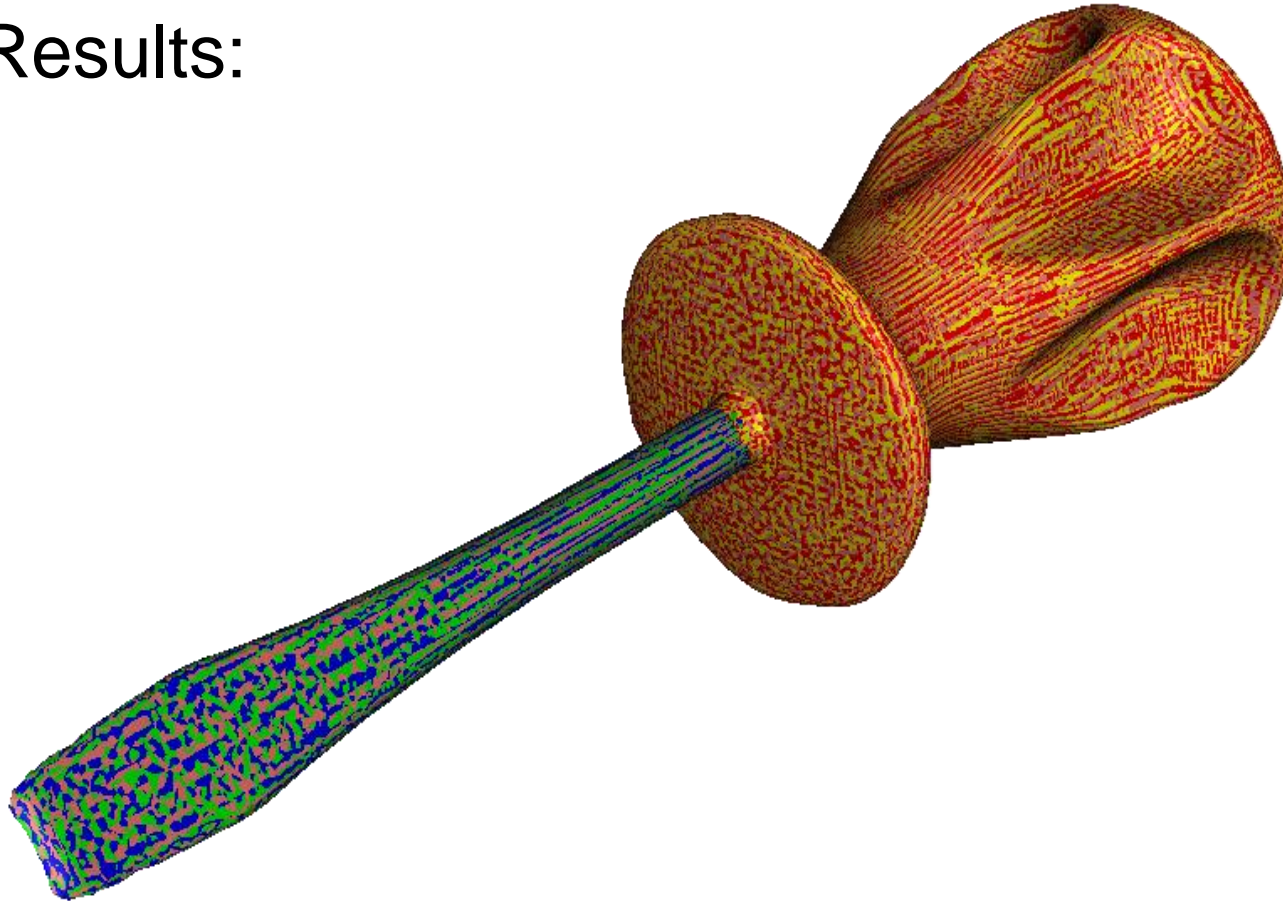


Principal
Symmetry Planes

Application: Beautification



Results:

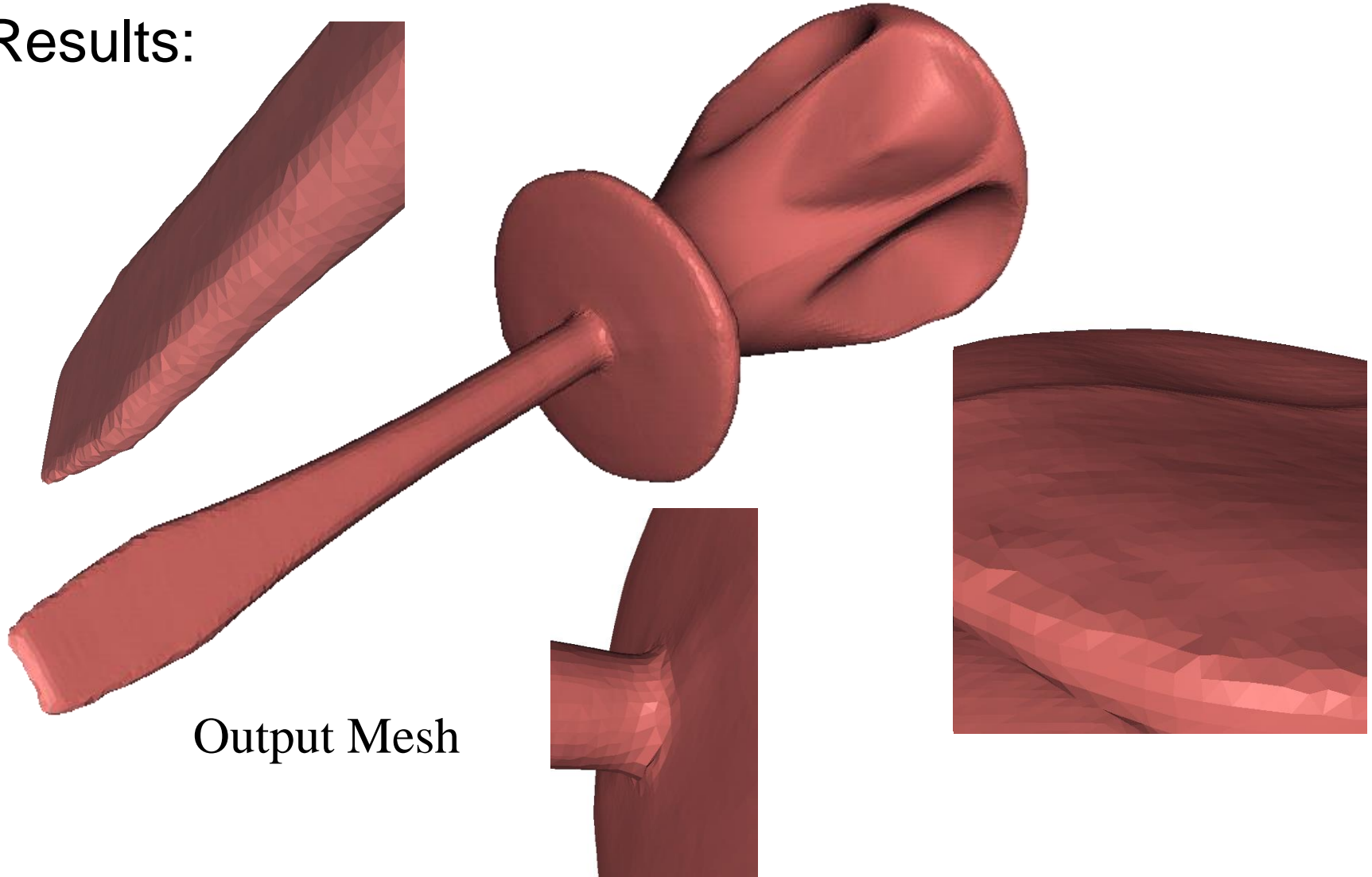


Output Mesh

Application: Beautification



Results:

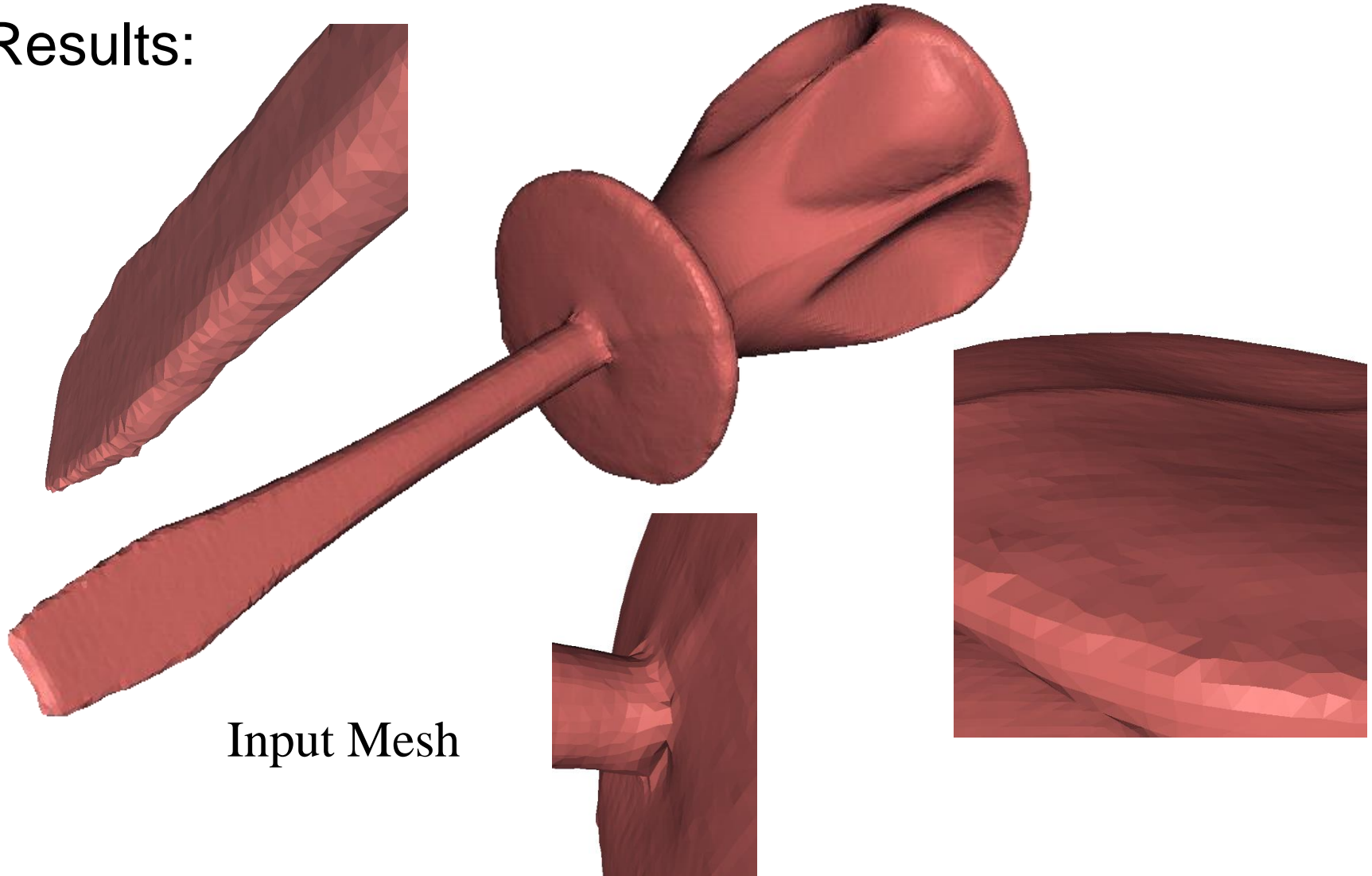


Output Mesh

Application: Beautification



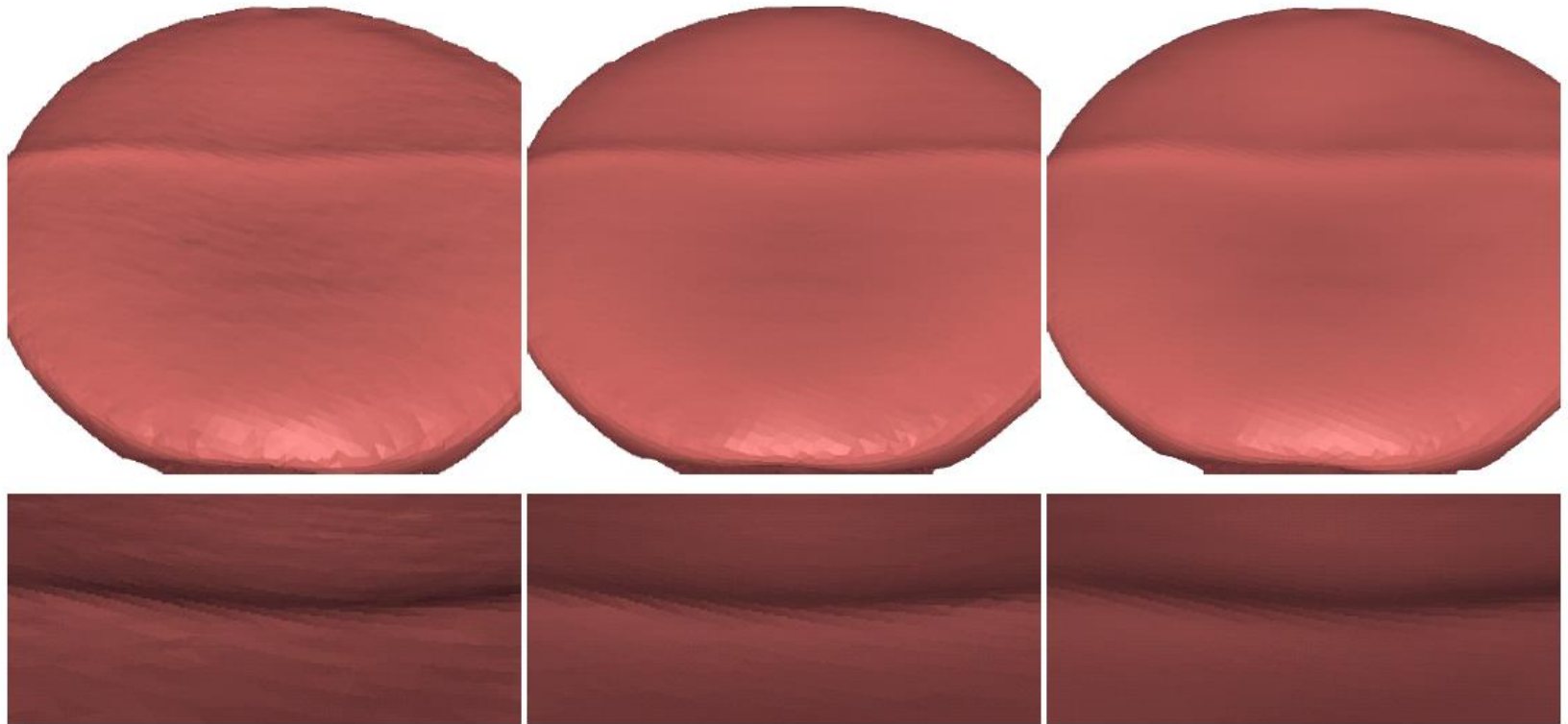
Results:



Application: Denoising



Results:



Input
Mesh

Symmetrized
Mesh

Bilateral
Filtering



Applications

Alignment

Matching

Segmentation

Viewpoint selection

Simplification

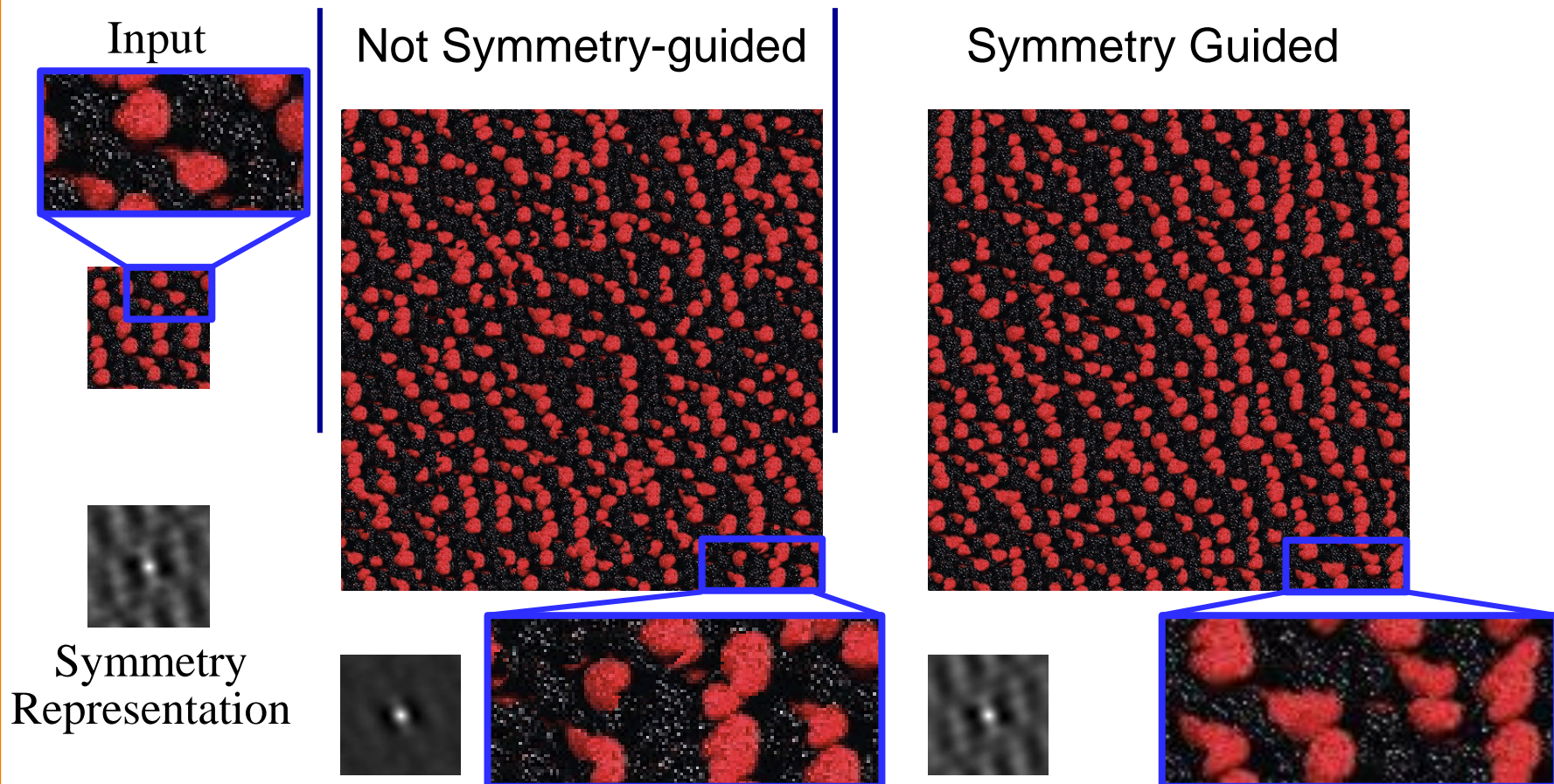
Beautification

Texture synthesis ←

Application: Texture Synthesis



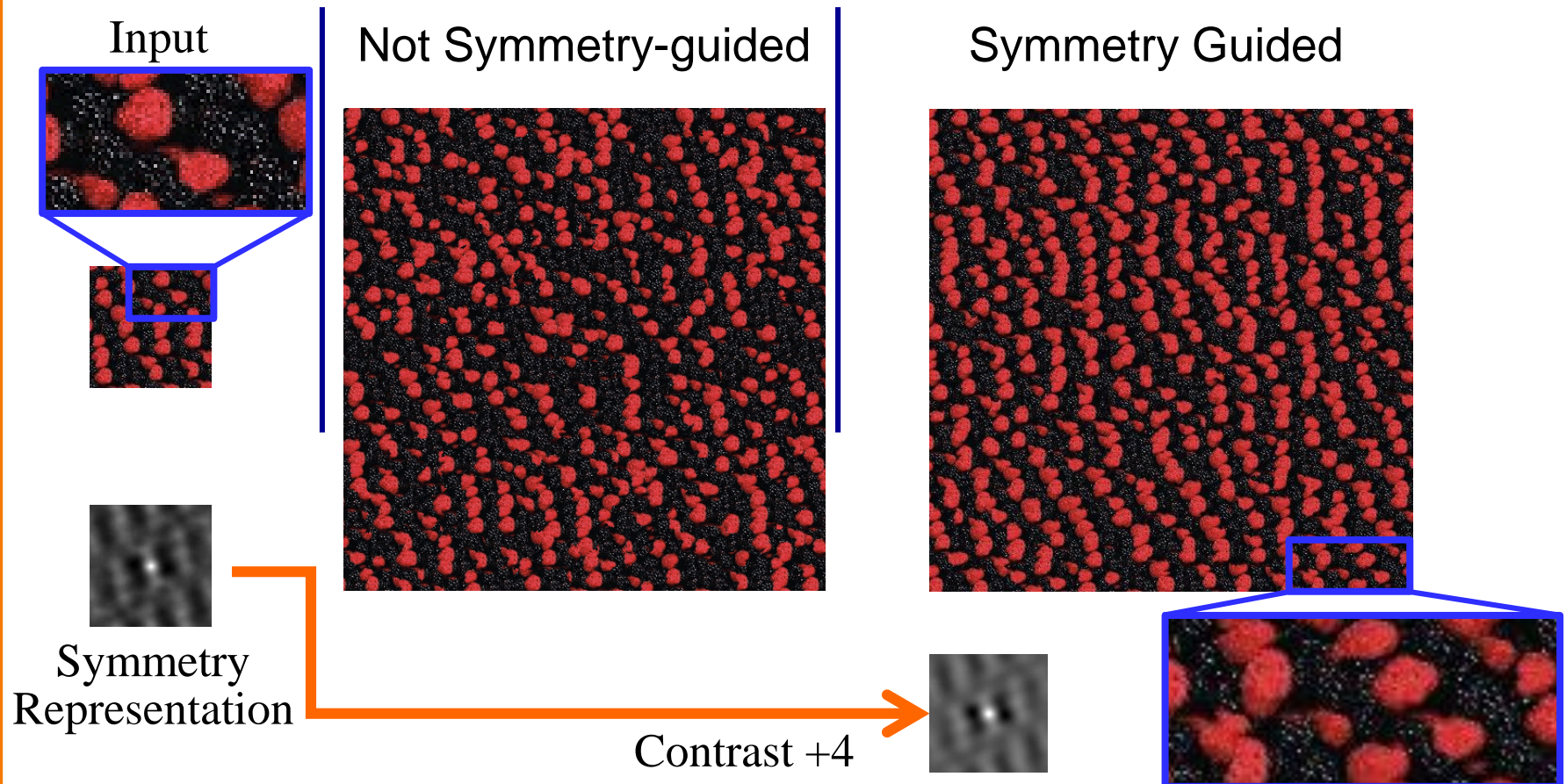
Maintain symmetries during texture synthesis



Application: Texture Manipulation



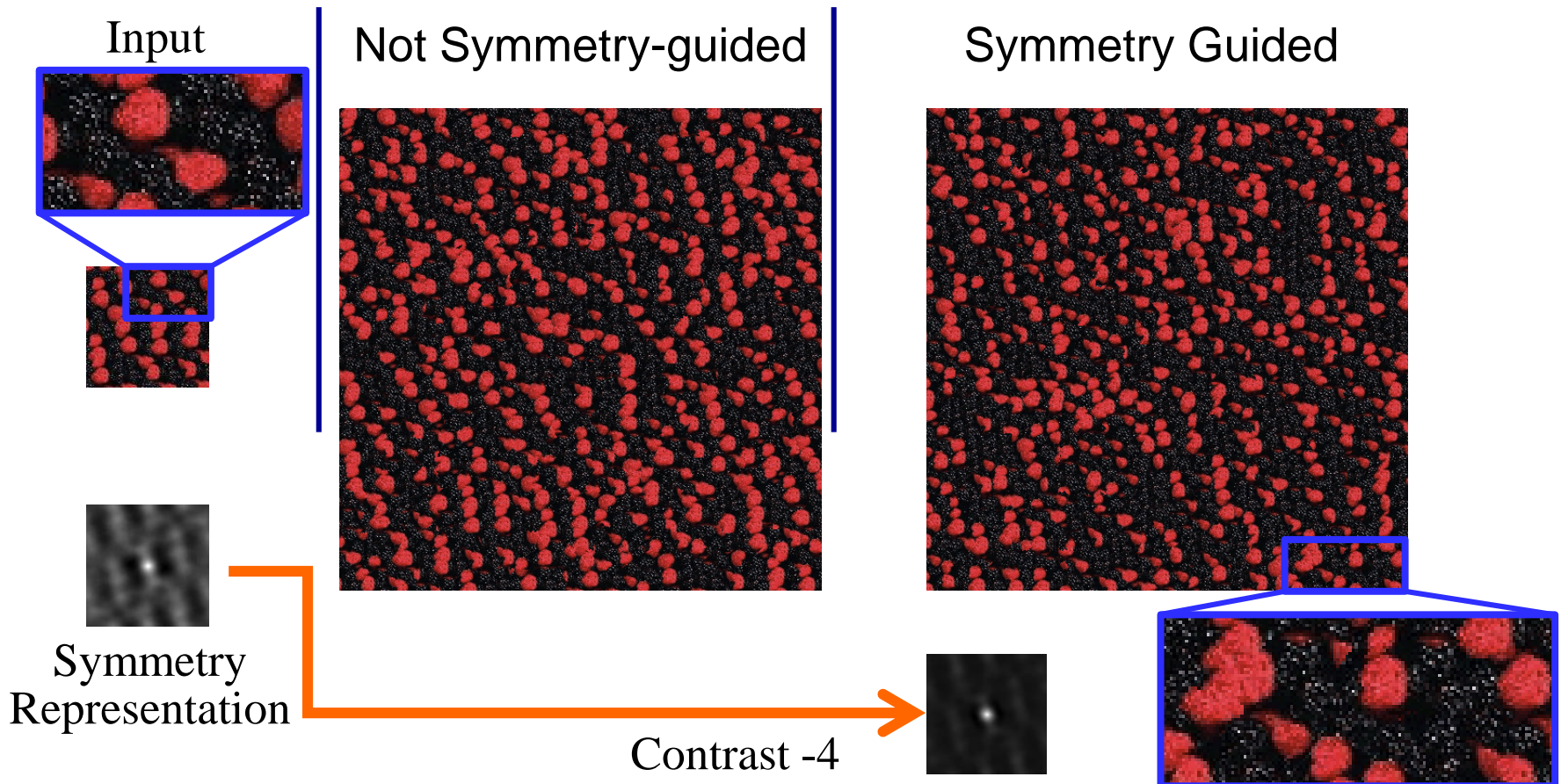
Edit symmetries during texture synthesis



Application: Texture Manipulation



Edit symmetries during texture synthesis



Summary



Representations

- Symmetry descriptor
- Symmetry transform
- Principal symmetries

Applications

- Alignment
- Matching
- Segmentation
- Viewpoint Selection
- Simplification
- Beautification
- Texture synthesis

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