

A Parametric Space Approach to the Computation of Multi-Scale Geometric Features

Anthousis Andreadis Georgios Papaioannou Pavlos Mavridis

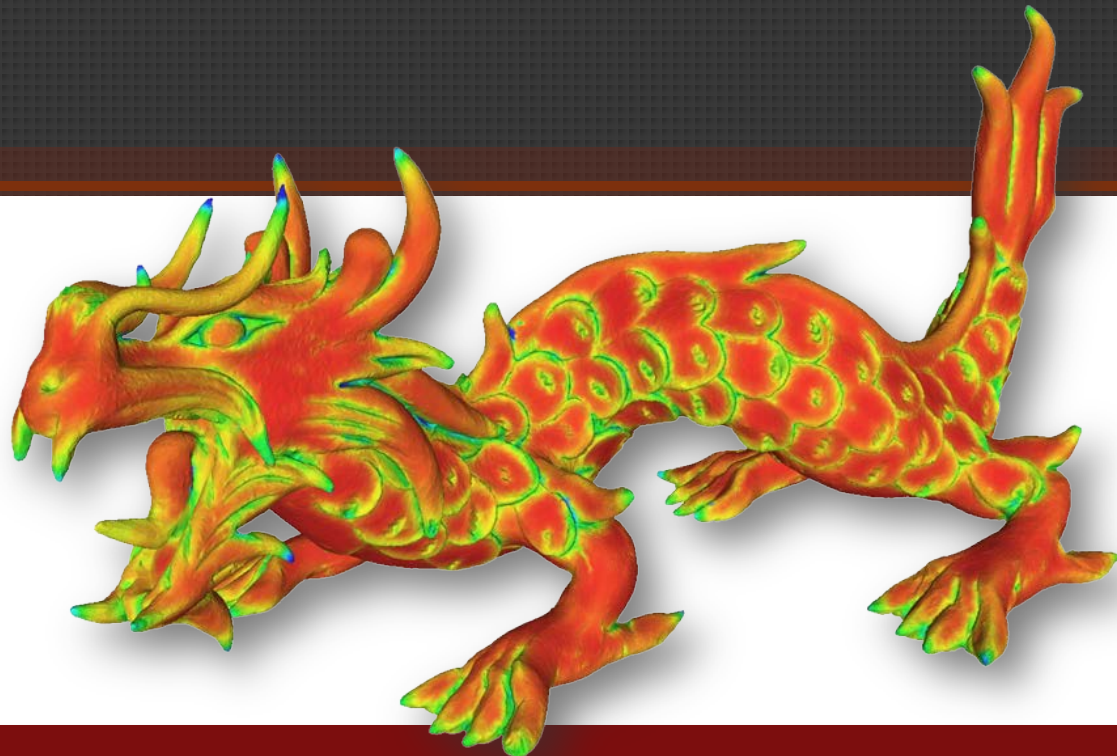
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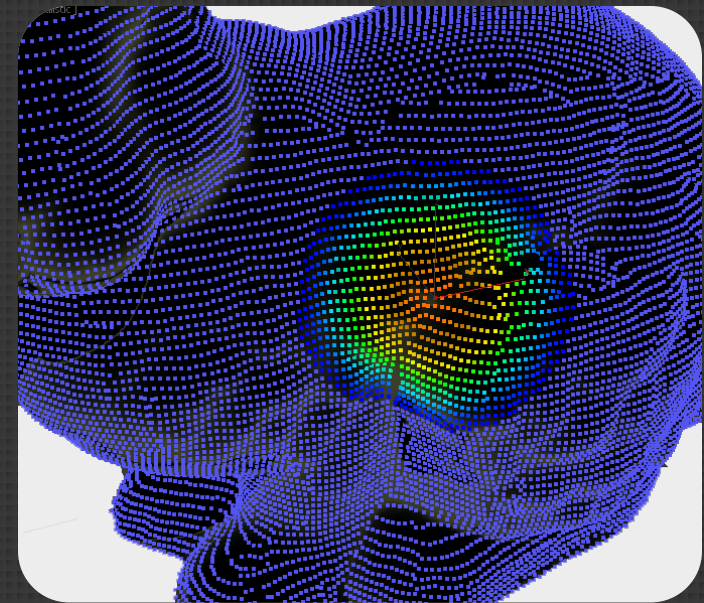
Introduction

- Geometric Features are central in a wide range of applications
 - Example Features:
Curvature, Shape Index
 - Example Applications:
Object Retrieval, Registration, Stylized Rendering
- Static geometry: Pre-compute
- Dynamic/Animated: Fast-computation is challenging



Feature Computation

- We focus on the general case of features with finite local support
- Key Element
 - Vertex Adjacencies/Point Neighbors
 - N-ring, Euclidean or Geodesic Distance





Related Work

- Existing methods can be classified based on the sampling method of the geometry
 - Object space
 - Volumetric
 - Screen Space
 - Parametric space

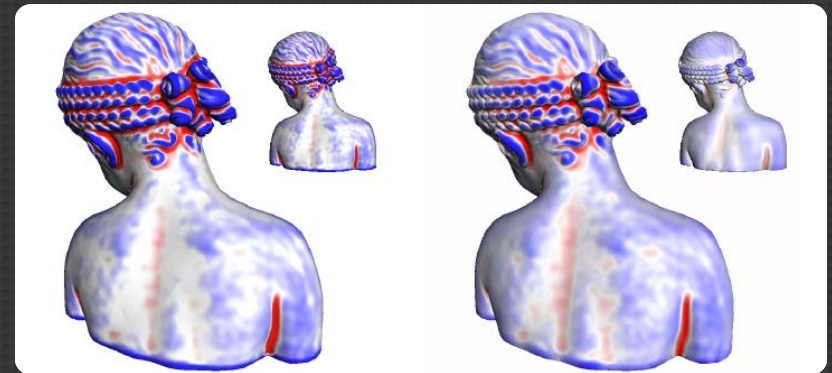


Object space

- Data structure encoding the adjacency is required (half-edge, kD-tree etc)
- These methods do not scale well as **computational complexity** is directly linked to
 - **Geometric density**
 - **Area of support**
- GPU mapping is non-trivial. Existing approaches do not generalize the sampling neighborhood. [Griffin et al., 2011]

Screen space

- Sample geometric information from a 2D pixel buffer.
- Adjacencies are implied by the pixel grid
 - Trivial sampling, efficient mapping to GPU's
- **Disadvantage:** Computations and area of support area limited to the visible point set
 - Inaccuracies, near occlusion points and at screen-space silhouettes

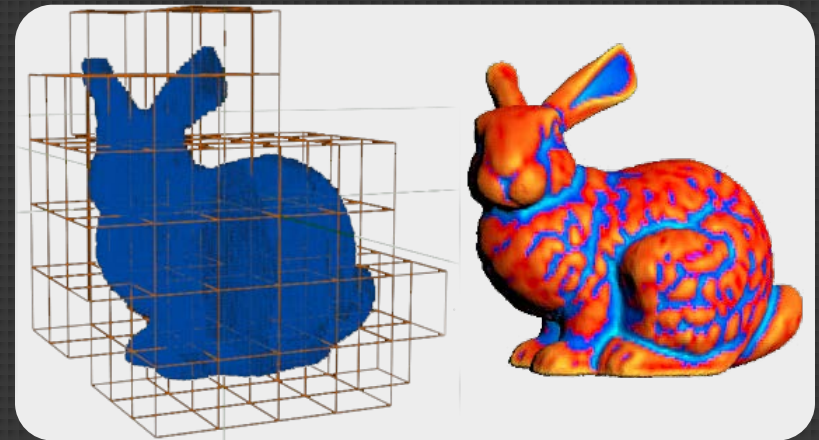


[Mellado et al., 2013]



Volumetric

- A volumetric representation is used (ex. level-set)
- Computational complexity now depends on the representation
- **Disadvantages**
 - Volumetric discretization is far more rough than the original surface
 - Incompatible results (ex. non-manifold surfaces)



[Museth, 2013]



Parametric Space

- Methods of this category rely on the unwrapped surface of the model on a 2D plane
- Computational **complexity decoupled from the geometry**
- **Disadvantages**
 - Neighbor discovery is not trivial
 - Cannot be performed directly on point clouds
- Existing methods are not generic
 - [Novatnack and Nishino, 2007] focus on image space techniques
 - [Hua et al.] Require specific unwrapping methodologies.



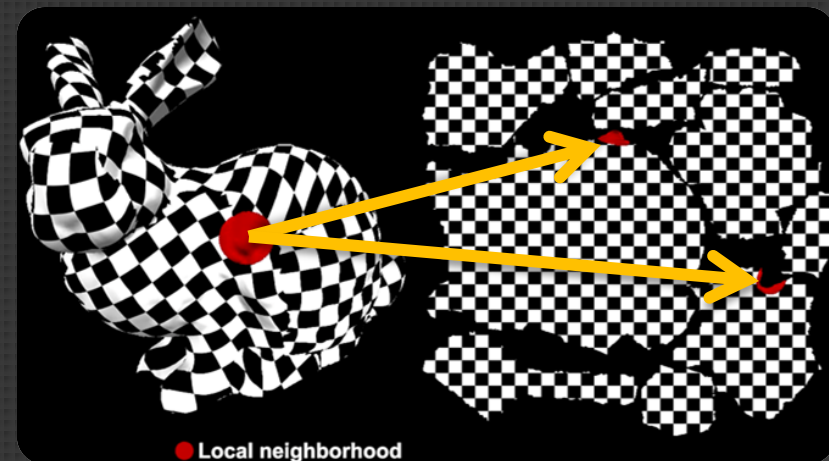
Motivation

- Design a method that is efficient, accurate and generic
 - **Efficiency**: Close to real-time even for large area of support for animated/deformable objects
 - **Excludes** Object Space
 - **Accuracy**: Similar results to a reference Object Space method
 - **Excludes** Screen Space & Volumetric
 - **Generality**: Not restricted to a specific feature, or parameterization



Method Overview

- Operates in parametric-space, but is agnostic to the actual mapping of the surface
- **Vertex Adjacencies → Pixel Adjacencies**
 - Not a perfect world: Chart boundaries create **discontinuities** of geometric adjacencies

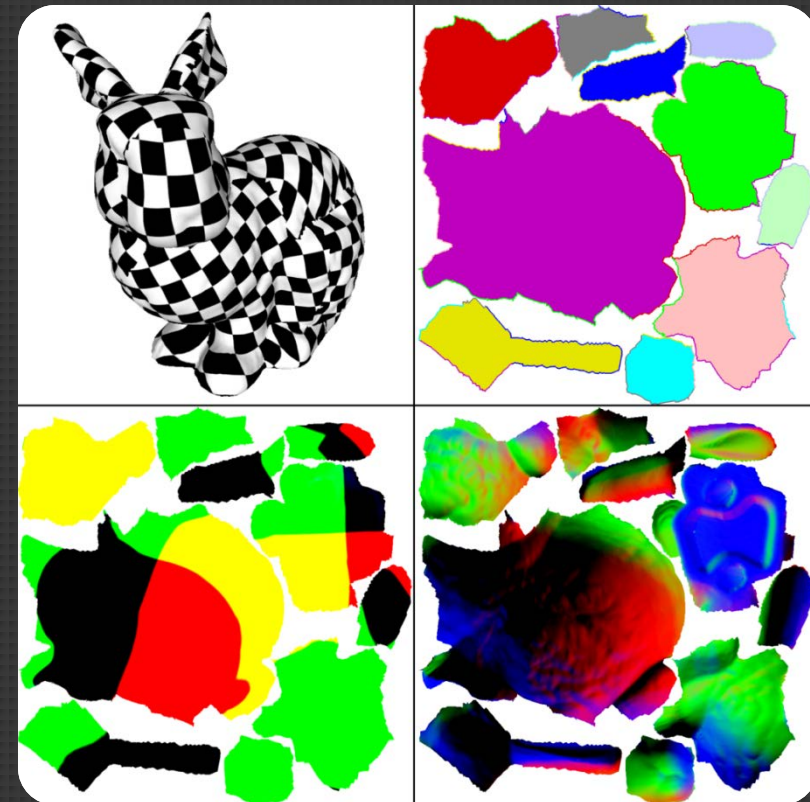




Method Overview

- **Pre-Process**
 - Locate affected edges and store extra information
- **Real-Time**
 - Create Data Buffers
 - Geometry, Normal, Adjacency
 - Recreate Adjacencies and perform Computations

Data Buffers





Data Buffers

- Information is stored in Textures
 - Geometry Buffer
 - *Object space positions, Chart id*
 - Normal Buffer
 - Adjacency Buffers
 - Discontinued Edges
 - *Adjacent chart id*
 - *Corresponding chart coordinates*
 - *Relative Scale & Rotation*
 - *Local metric distortion (LMD)*
 - *Angular distortion*
 - *u, v stretch factors*





Data Buffer Generation

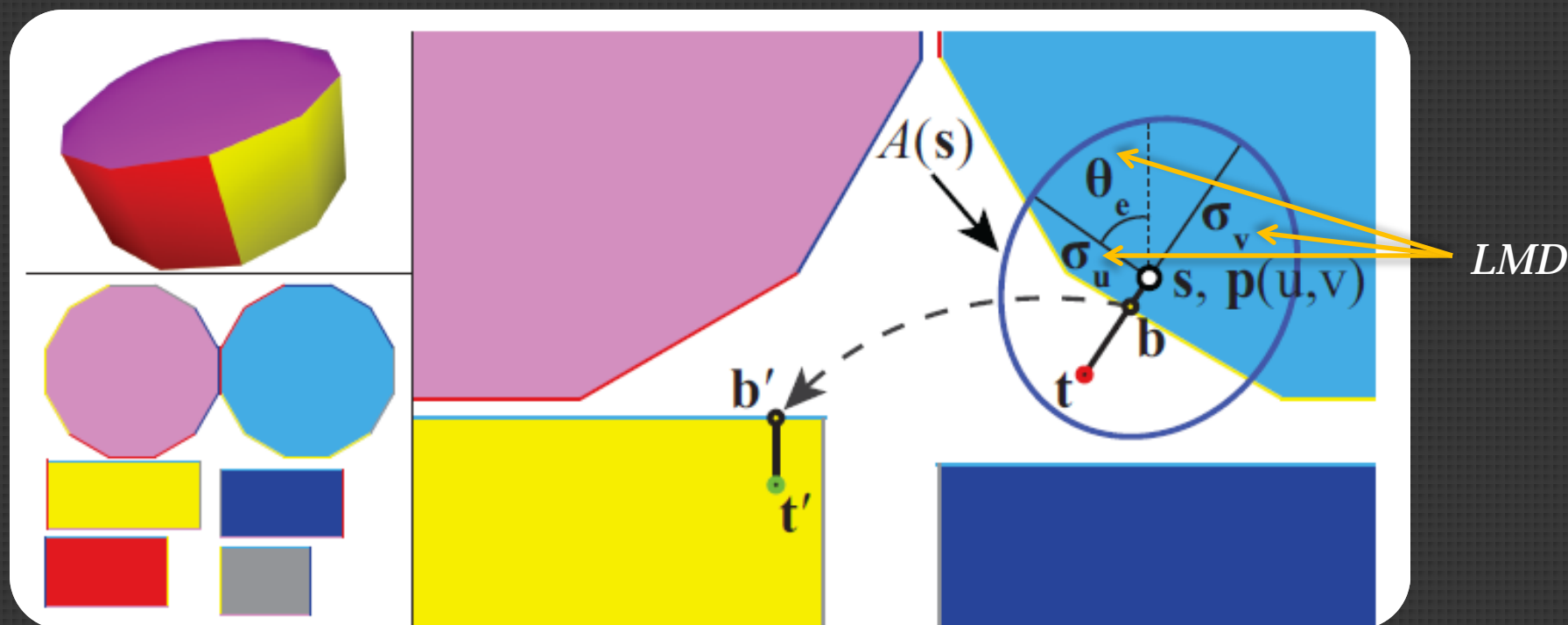
- Rasterize object triangles
 - Chart boundary edges are rasterized separately to avoid disconnected regions
- LMD factors computed using eigen-decomposition of the *first fundamental form matrix*
 - Used for the **anisotropic adjustment of scale and sampling directions**

$$J_P^T J_P = \begin{bmatrix} E & F \\ F & G \end{bmatrix} \quad \begin{aligned} E &= (\partial P(u, v) / \partial u)^2 & G &= (\partial P(u, v) / \partial v)^2 \\ F &= (\partial P(u, v) / \partial u) \cdot (\partial P(u, v) / \partial v) \end{aligned}$$



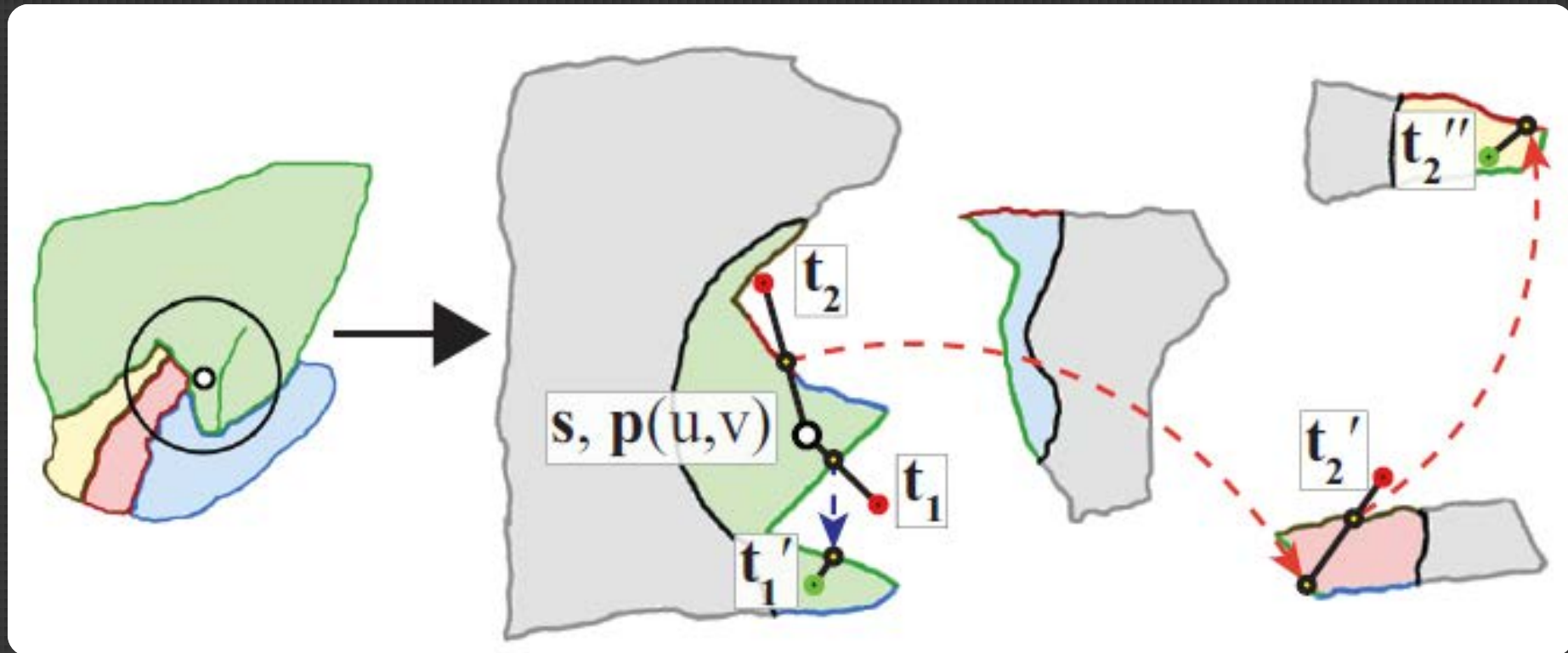
Sampling the Neighborhood of a Point

$$\mathbf{t}' = \mathbf{b}' + \mathbf{R}_{\theta(\mathbf{b} \rightarrow \mathbf{b}')} \mathbf{S}_{s(\mathbf{b} \rightarrow \mathbf{b}')} (\mathbf{t} - \mathbf{b}) \quad s(\mathbf{b} \rightarrow \mathbf{b}') = \left(\frac{\sigma_u(\mathbf{b}')}{\sigma_u(\mathbf{b})}, \frac{\sigma_v(\mathbf{b}')}{\sigma_v(\mathbf{b})} \right)$$





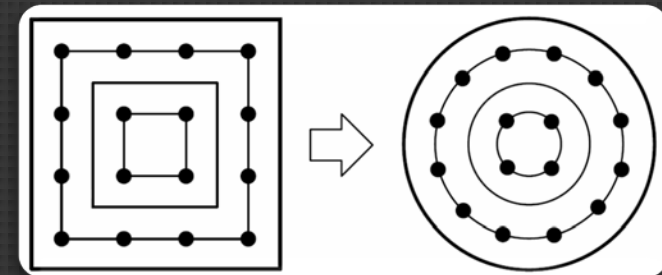
Sampling the Neighborhood of a Point





Monte Carlo Integration

- Geometric feature computation is usually performed with surface and volume integrals
- We estimate by Monte Carlo integration
- Generate random samples using a stratification scheme on a grid and transform them to disk using concentric mapping
- Disk samples are anisotropically scaled and rotated according to *LMD* factors.
- We sample $A(s)$ ellipse using sample rejection based on the criterion of neighborhood $S(p)$

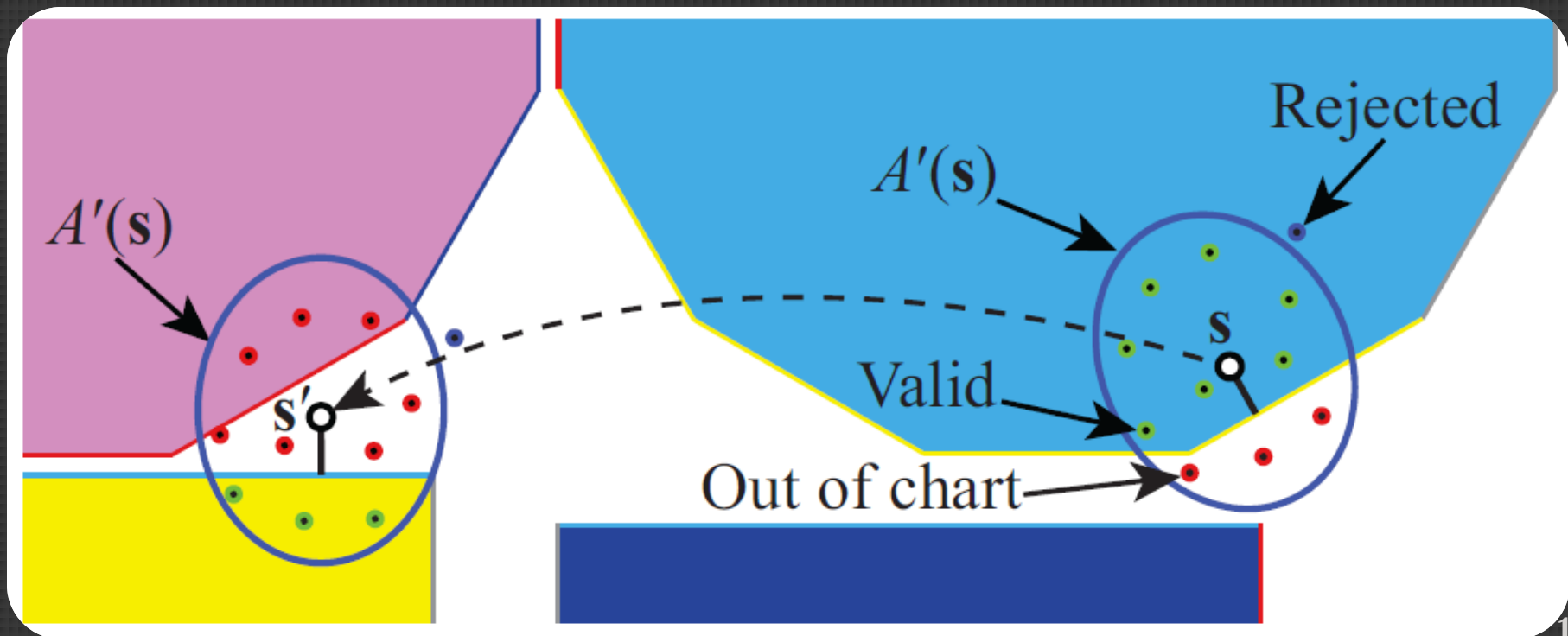


[Shirley and Chiu, 1997]



Monte Carlo Integration

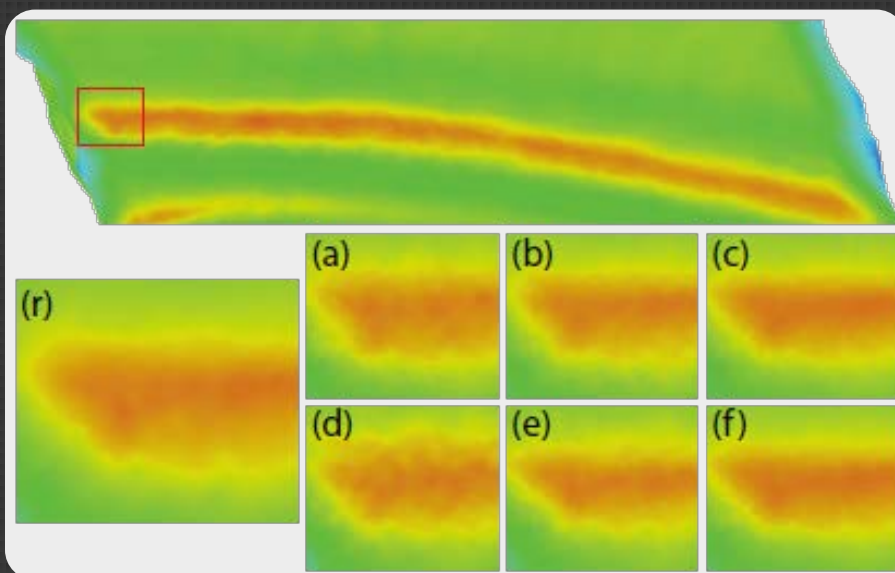
$$\langle I \rangle(\mathbf{p}) = \frac{A'(\mathbf{s})}{N} \sum_{i=1}^N g(P(\mathbf{t}_i))$$





Adaptive Sampling

- Smooth surface areas converge faster than areas with high variance
- We use simplified two-step adaptive sampling



Samples	Full		Adaptive	
	Time	% AE	Time	% AE
64	(a) 17.57ms	1.172	(d) 15.94ms	1.331
100	(b) 22.17ms	1.035	(e) 19.54ms	1.110
256	(c) 50.54ms	1.005	(f) 41.44ms	1.007



Results

- Implemented Geometric Features
 - Mean Curvature
 - Local Bending Energy
 - Normalized Sphere Volume
 - Shape Index
- Comparison with multi-core CPU object space approach using Half-Edge data structure

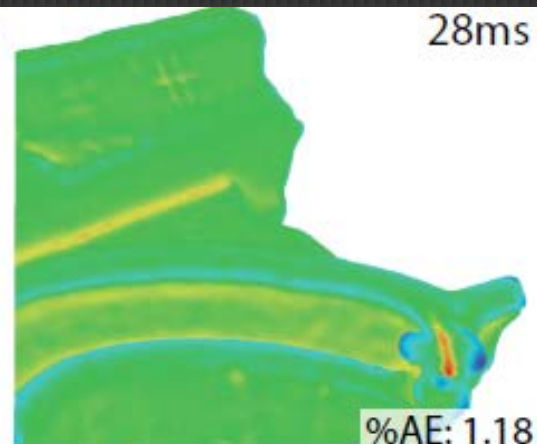
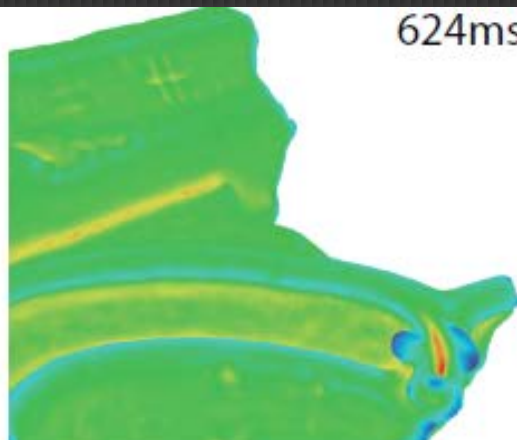


Results (Mean Curvature)

Reference

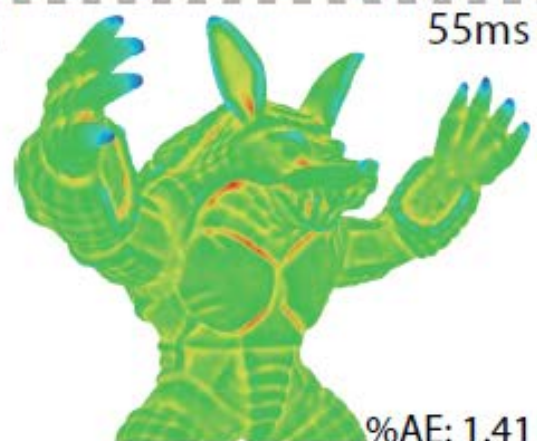
Our Method

Embrasure
200K Triangles
340x334x330mm
10mm Radius



~22x

Armadillo
345K Triangles
126x115x152mm
3mm Radius



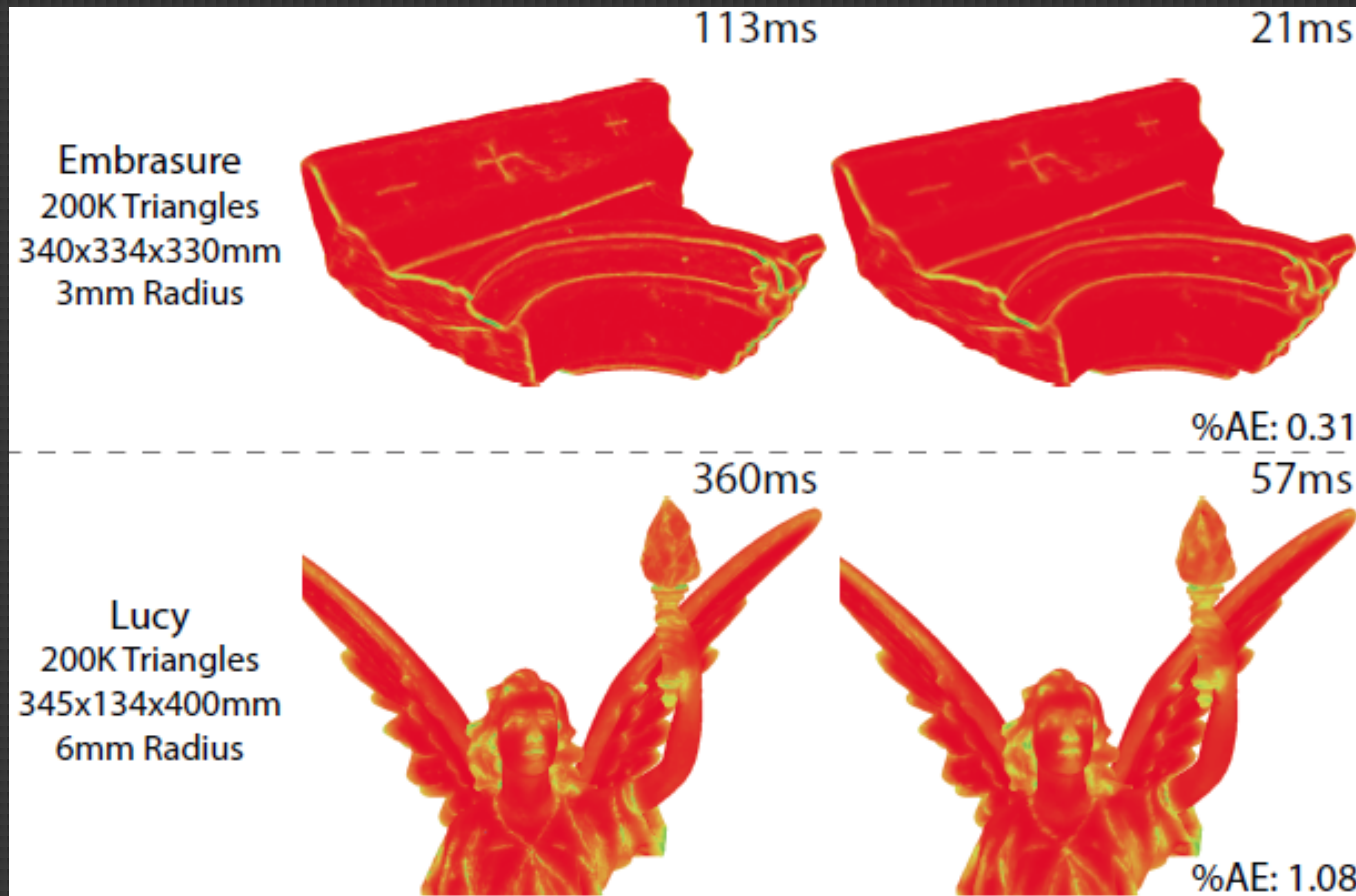
~25x



Results (Local Bending Energy)

Reference

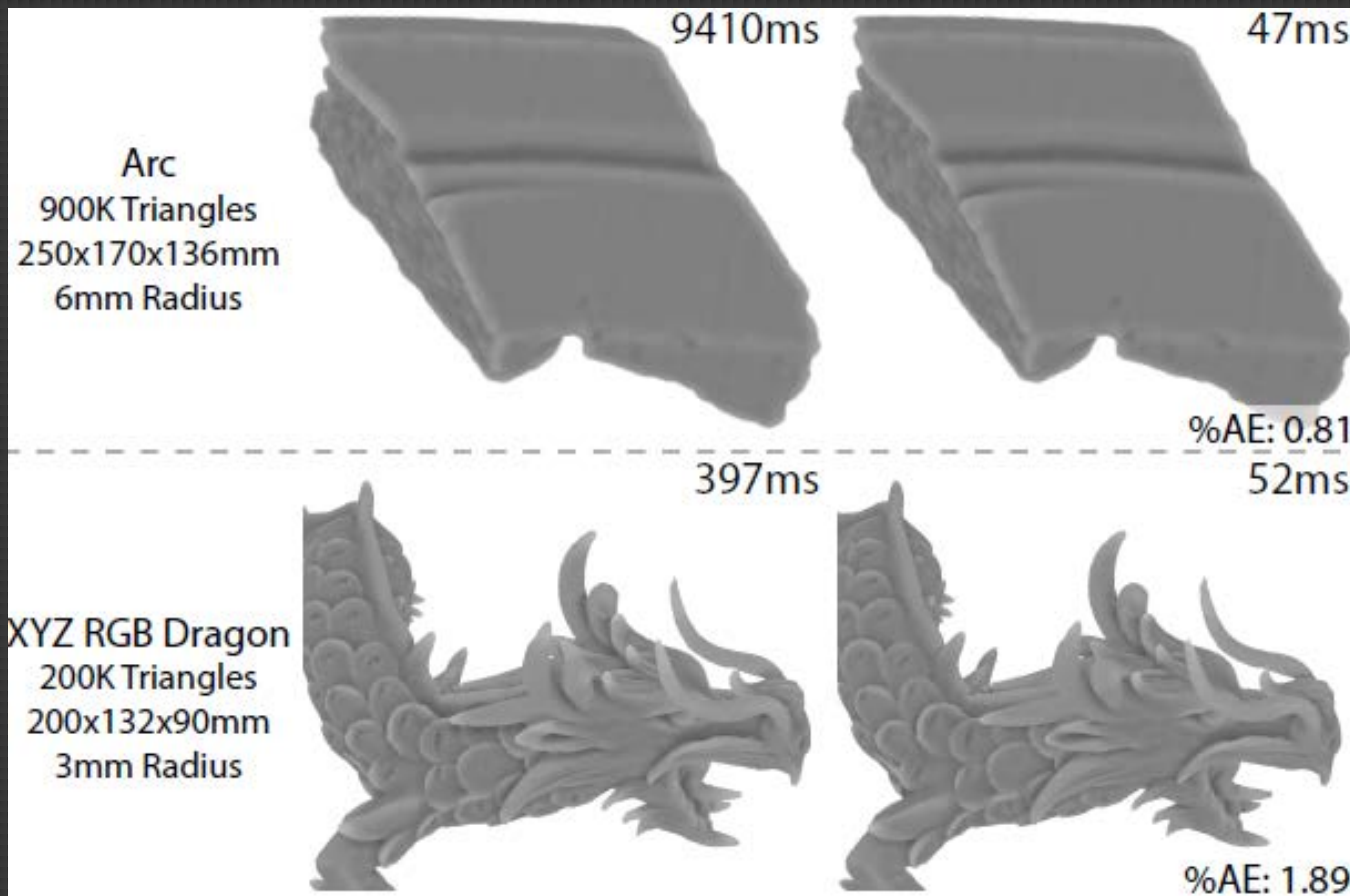
Our Method



Results (Sphere Volume)

Reference

Our Method

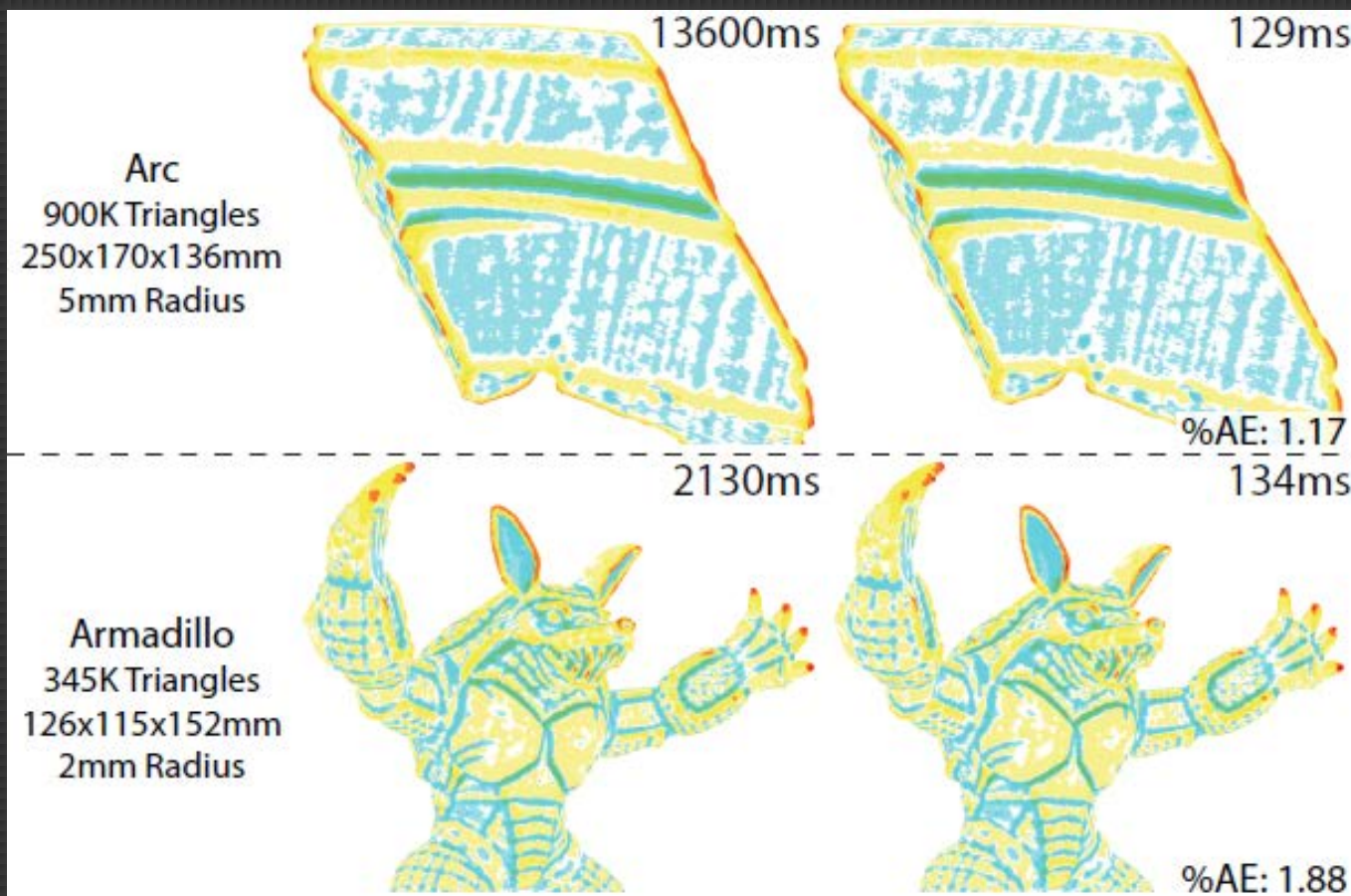




Results (Shape Index)

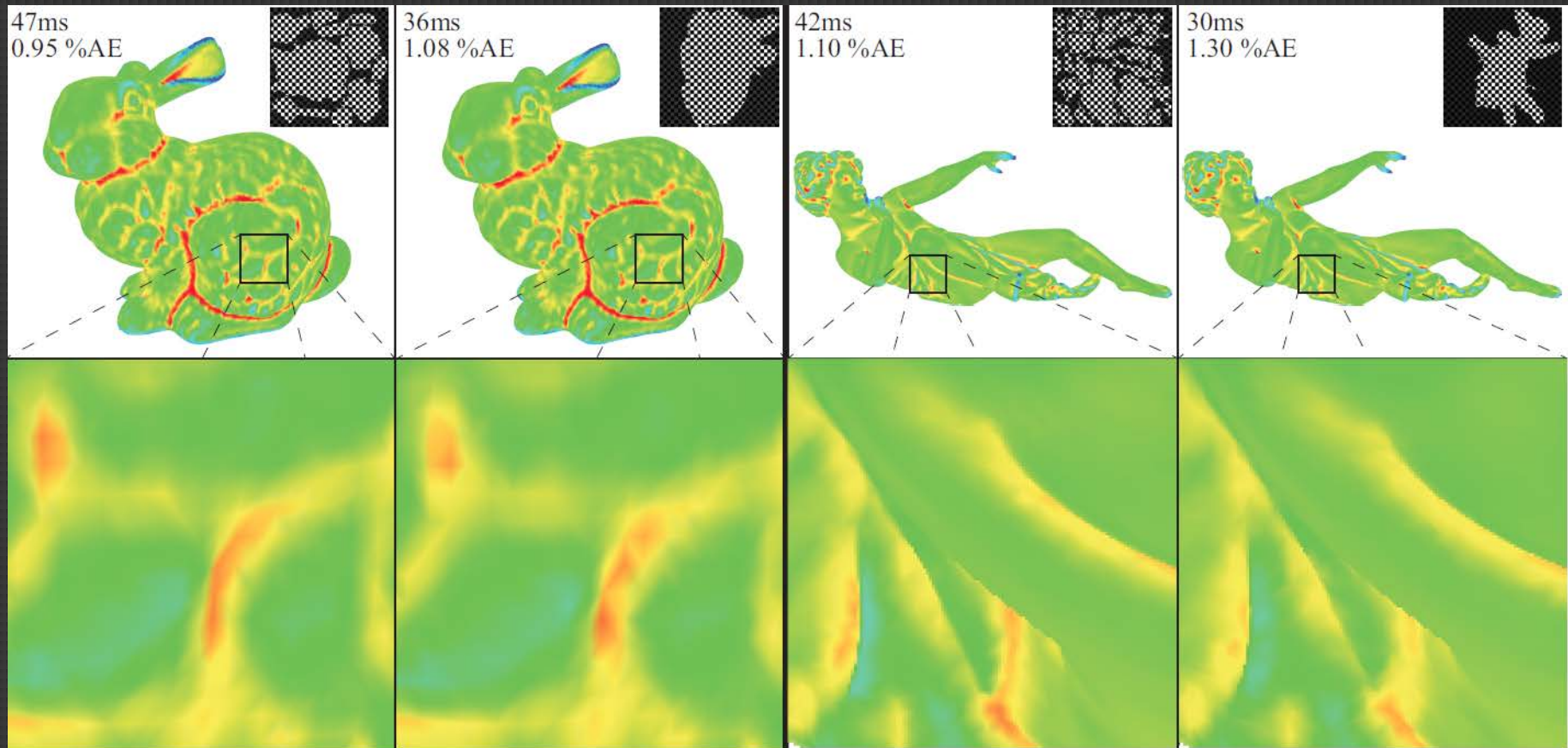
Reference

Our Method



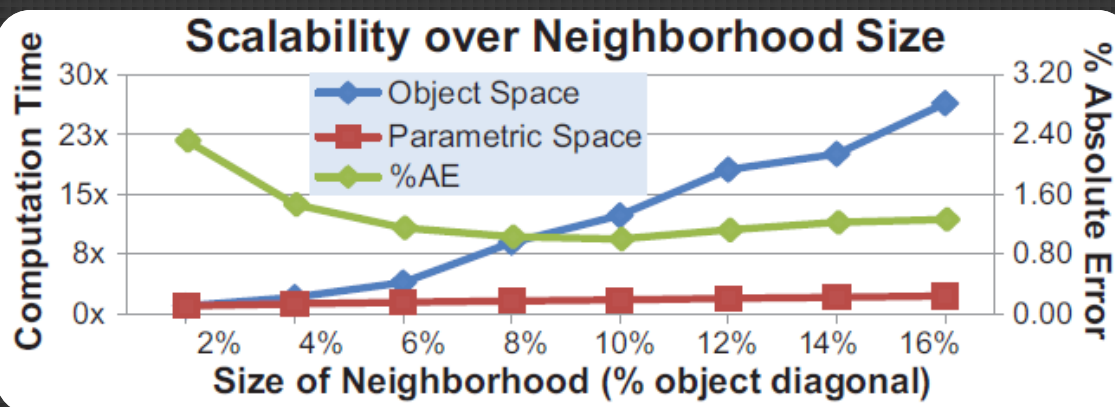
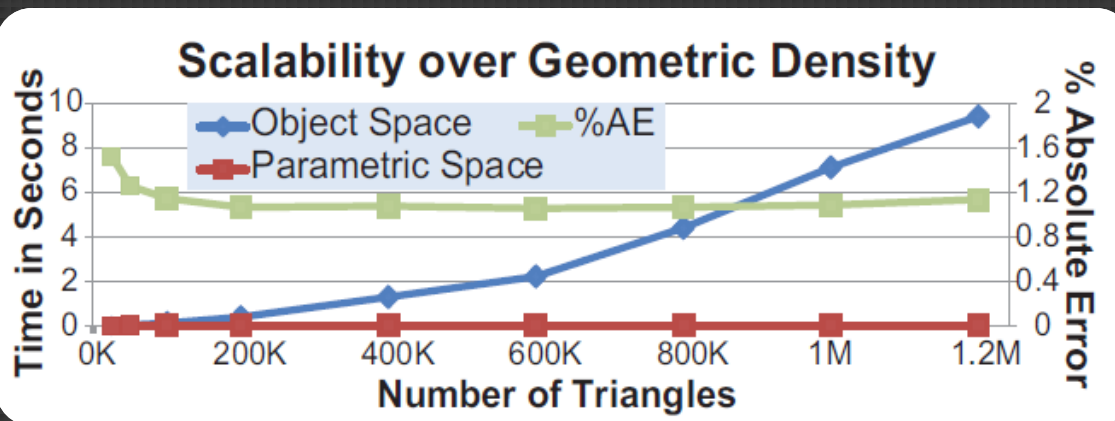


Results (Different Parameterizations)





Results (Scalability)





Thank you!

- Questions ?
- More info:
 - <http://presious.eu>
 - <http://graphics.cs.aueb.gr>