

# ALICE

## Bruno Lévy

INSTITUT NATIONAL  
DE RECHERCHE  
EN INFORMATIQUE  
ET EN AUTOMATIQUE



*Evaluation seminar, October 21, 22*

*Theme: Interaction and Visualization*

# Overview

- **Introduction**
  - Overall objectives
  - Composition of the team
- **Zoom on Geometry Processing**
  - Fitting and Parameterization
  - Sampling and Meshing
- **Impact**
- **Evolution and Future Work**



# Introduction

Overall objectives - geometry and light



Light



# Introduction

Overall objectives - geometry and light

## *Light*

- Realistic rendering
- Interactive rendering
- Scientific visualization



Light



# Introduction

Overall objectives - geometry and light



Geometry



Light



# Introduction

Overall objectives - geometry and light



## *Geometry*

- Optimizing...
- Transforming...
- Constructing...

...representations

Geometry



# Composition of the team

**May 2006 (creation):**

**4 permanent researchers**





# Composition of the team

## *May 2006 (creation):*

4 permanent researchers

## *October 2010:*

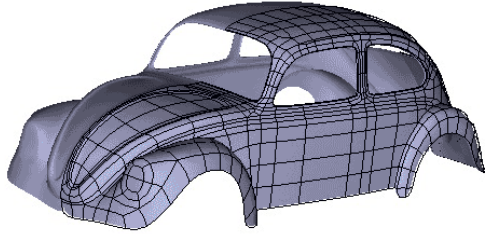
7 permanent researchers

1 visiting associate professor





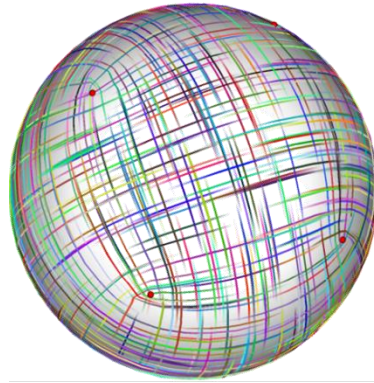
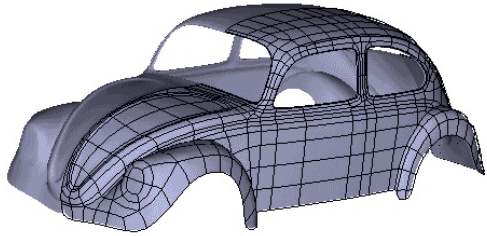
# Zoom on Geometry Processing Overview



## 1. Intro Dynamic Function Basis



# Zoom on Geometry Processing Overview

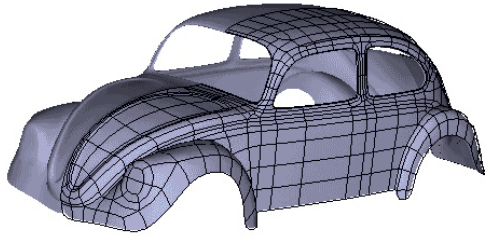


**1. Intro**  
**Dynamic Function Basis**

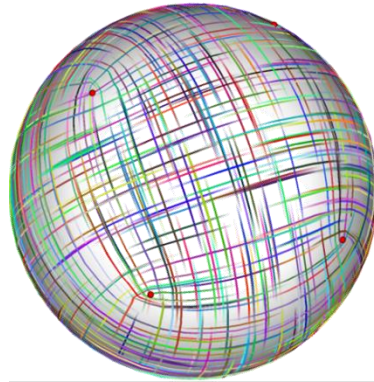
**2. Direction Fields**



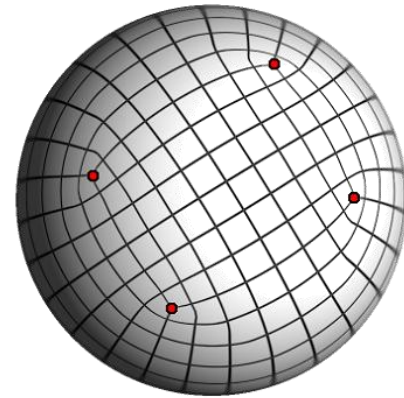
# Zoom on Geometry Processing Overview



**1. Intro**  
**Dynamic Function Basis**



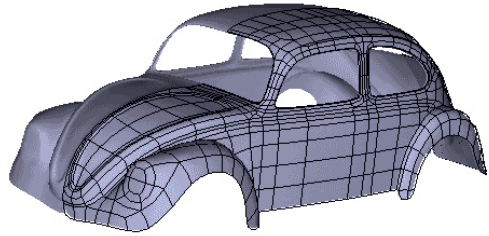
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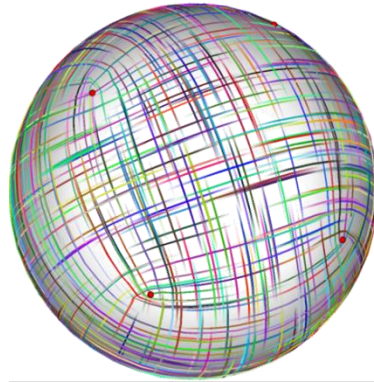
**3. Global Parameterization**



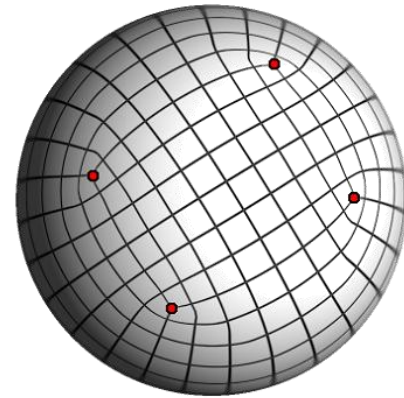
# Zoom on Geometry Processing Overview



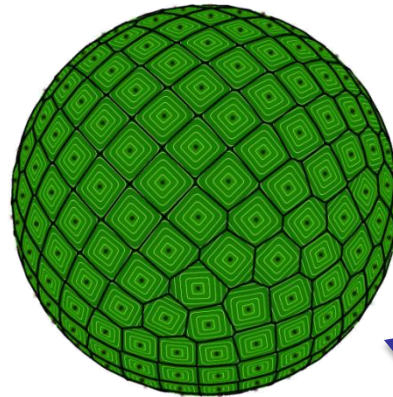
1. Intro  
Dynamic Function Basis



2. Direction Fields



3. Global Parameterization

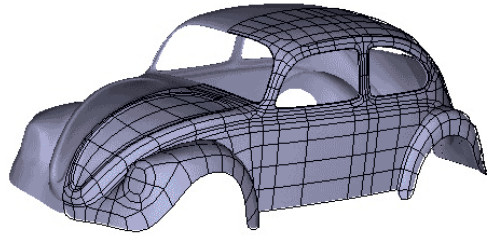


4. Optimal Sampling ( $L_p$  and  $L_2$ )

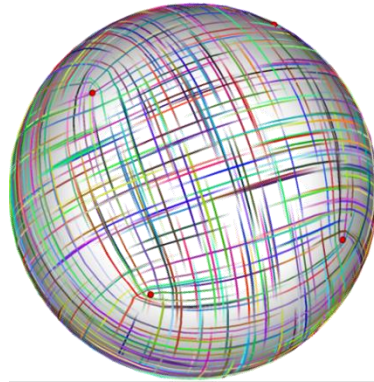




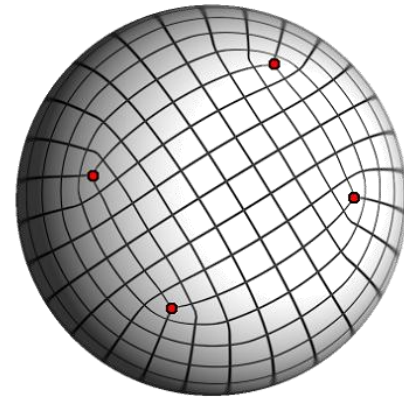
# Zoom on Geometry Processing Overview



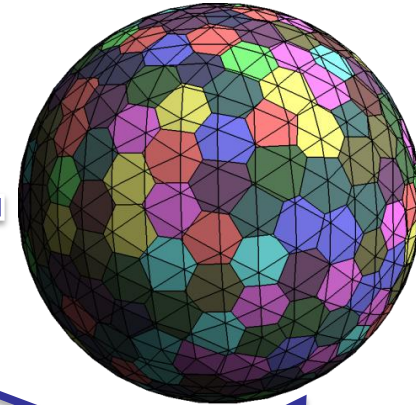
1. Intro  
Dynamic Function Basis



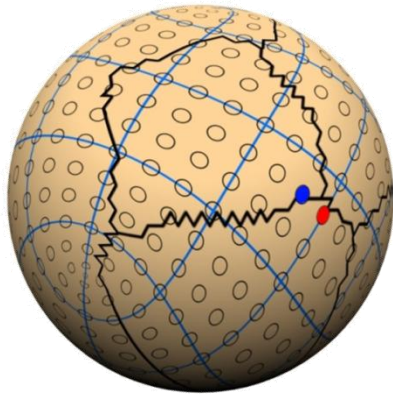
2. Direction Fields



3. Global Parameterization



4. Optimal Sampling ( $L_p$  and  $L_2$ )

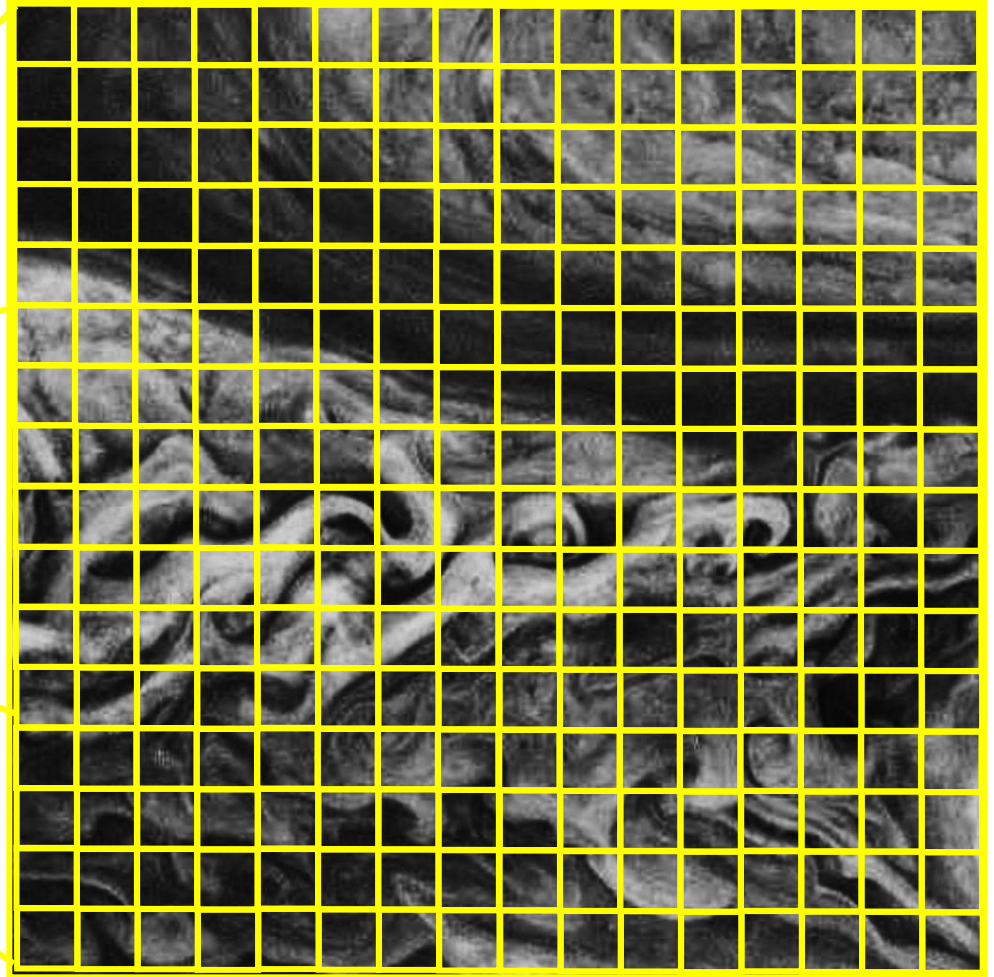
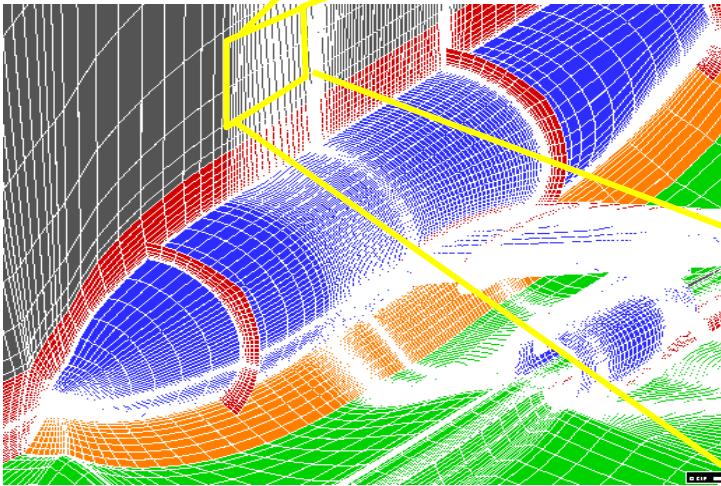


5. Seamless Texturing



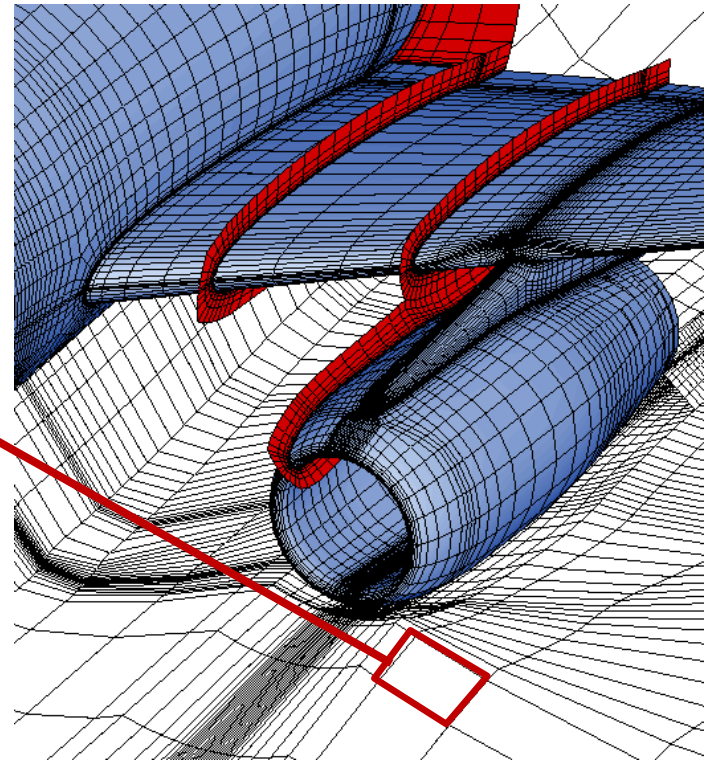
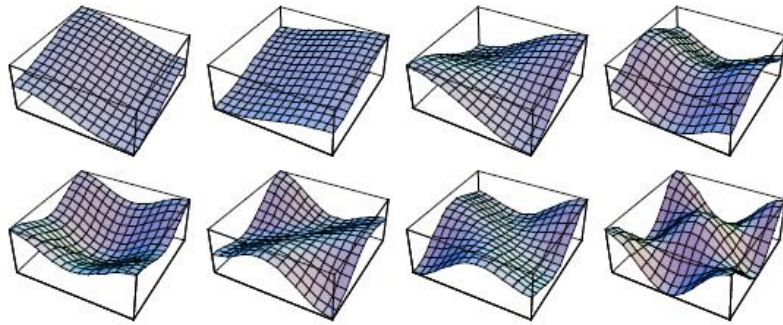
# Zoom on Geometry Processing

## 1. Dynamic Function Basis – classical FEM



# Zoom on Geometry Processing

## 1. Dynamic Function Basis – classical FEM



•Function basis ( $\phi_i$ ):  $f = \sum \alpha_i \phi_i$

•Operator equation:  $Lf = g$

•Hilbert space, Inner Product:  $\langle f, g \rangle = \int f(x) g(x) dx$

• $\forall i, \langle Lf, \phi_i \rangle = \langle g, \phi_i \rangle$





# Zoom on Geometry Processing

## 1. Dynamic Function Basis – New framework

$$f = \sum \alpha_i \phi_i(\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_m, \mathbf{x}, \mathbf{y})$$

$$= \sum \alpha_i \phi_i(\mathbf{p}, \mathbf{x})$$



# Zoom on Geometry Processing

## 1. Dynamic Function Basis – New framework

$$f = \sum \alpha_i \phi_i(\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_m, \mathbf{x}, \mathbf{y})$$

$$= \sum \alpha_i \phi_i(\mathbf{p}, \mathbf{x})$$

Galerkin:  $\forall i, \langle Lf, \phi_i \rangle = \langle g, \phi_i \rangle$



# Zoom on Geometry Processing

## 1. Dynamic Function Basis – New framework

$$f = \sum \alpha_i \phi_i(\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_m, \mathbf{x}, \mathbf{y})$$

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Galerkin:  $\forall i, \langle Lf, \phi_i \rangle = \langle g, \phi_i \rangle$

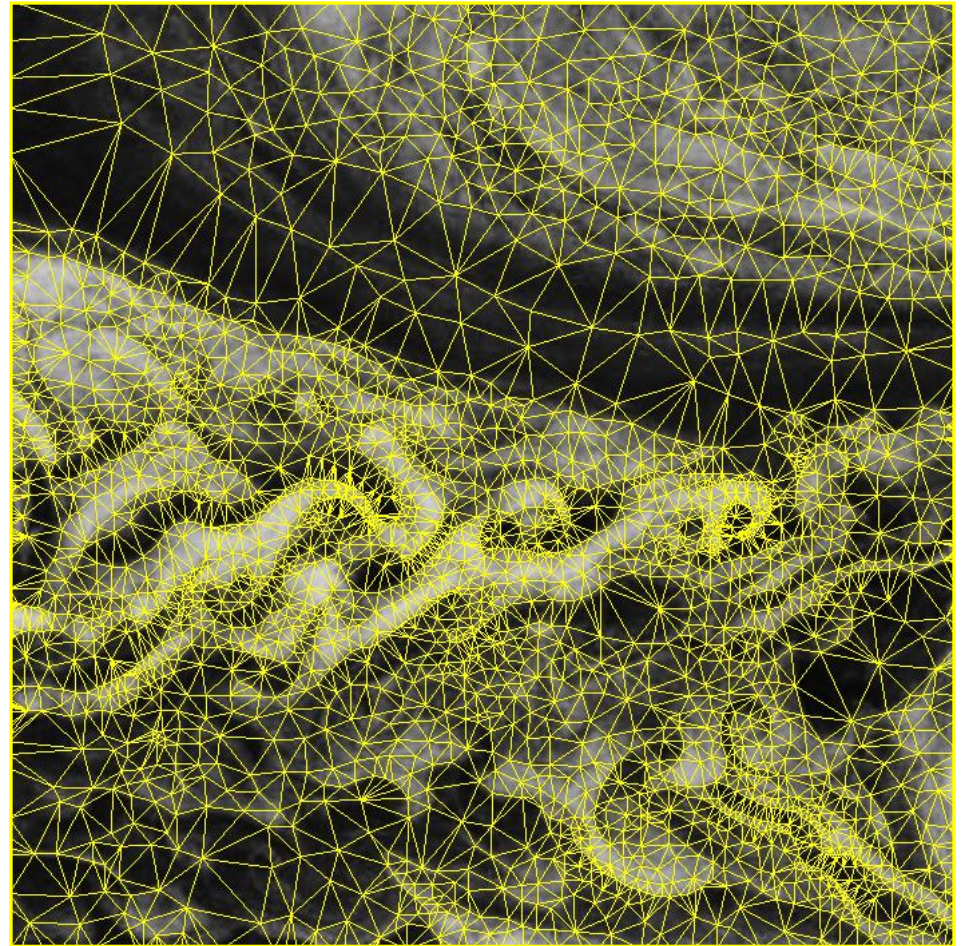
DFB: minimize  $F(\mathbf{p}, \alpha) = |Lf - g|^2 = \left| \sum \alpha_i \phi_i(\mathbf{p}, \mathbf{x}) - g \right|^2$

Solve for  $f$  [ $\alpha$ ] and for its sampling [ $\mathbf{p}$ ]



# Zoom on Geometry Processing

## 1. Dynamic Function Basis – Expected result



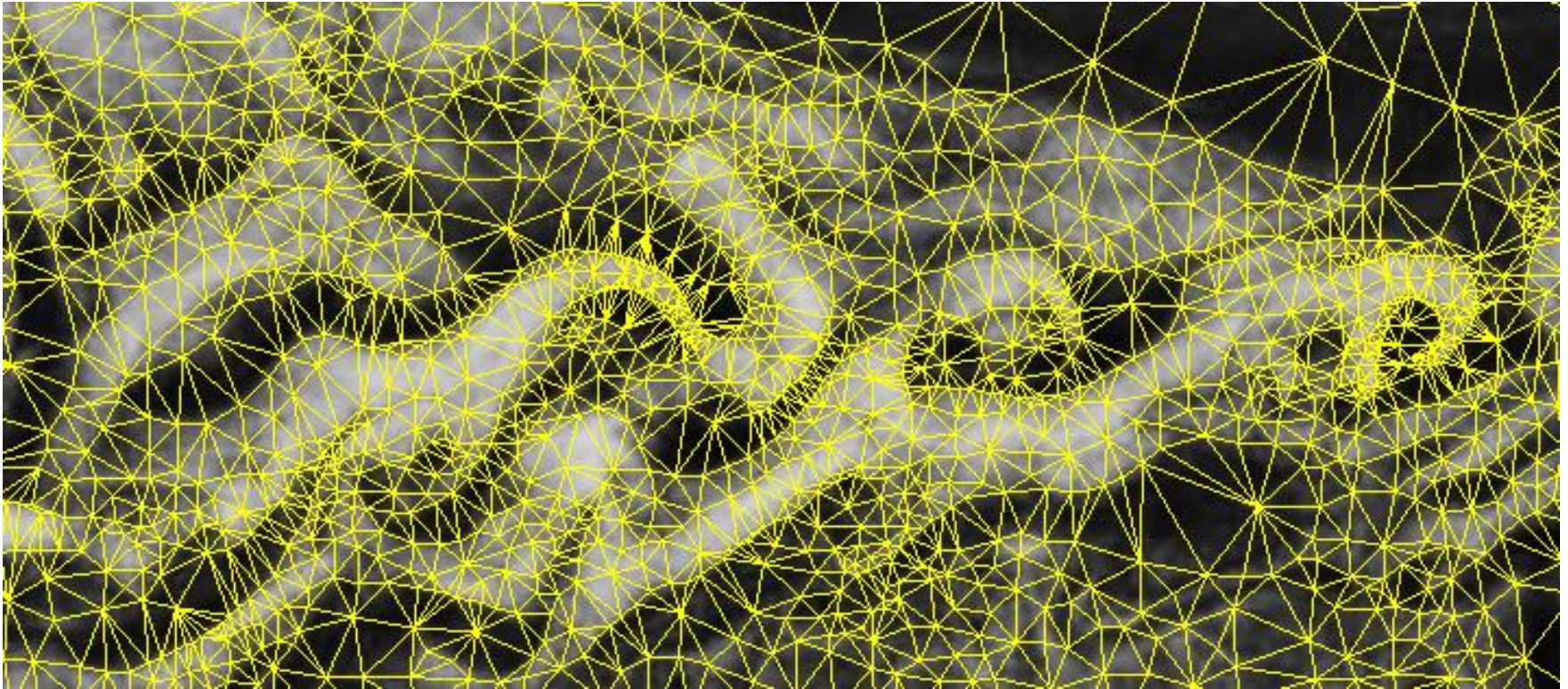
Our new framework:  
Dynamic Function Basis  
(DFB)





# Zoom on Geometry Processing

## 1. Dynamic Function Basis – Expected result



Our new framework: Dynamic Function Basis  
Solve for **approximation** and **sampling** all together



# Zoom on Geometry Processing

## Dynamic Function Basis – Research Program

Geometric Intelligence  
*Microsoft Research*

GOODSHAPE  
*European Research Council*  
1.1 Meuros, 5 years  
0.3% acceptance  
*all disciplines of science*

- 2D,  $L = Id$  : image approximation **[EGSR 2006]**
- 3D,  $L = Id$  : surface approximation **2006-2010**
- 3D, optimal sampling **2006-2010**
- 3D,  $L =$  light transport **2010-...**
- 3D+t, Navier Stokes **2010-...**

$$Lf = g$$



# Zoom on Geometry Processing

## Dynamic Function Basis – Research Program



- 2D,  $L = Id$  : image approximation **[EGSR 2006]**
- 3D,  $L = Id$  : surface approximation **2006-2010**
- 3D, optimal sampling **2006-2010**
- 3D,  $L =$  light transport **2010-...**
- 3D+t, Navier Stokes, tracking **2010-...**

$$f = g$$





# Zoom on Geometry Processing

## 1. Surface approximation – the challenge

Creating a CAD model from a real car ...



1970's: purely manual acquisition (Y. Sutherland)



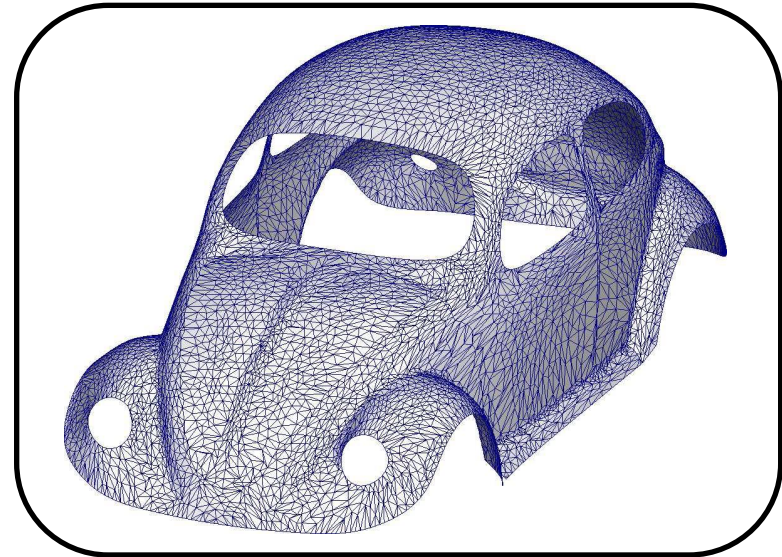
# Zoom on Geometry Processing

## 1. Surface approximation – the challenge

Creating a CAD/CAM model of a car



3D laser scanner



Reconstructed shape

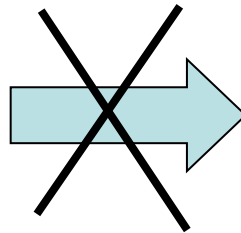
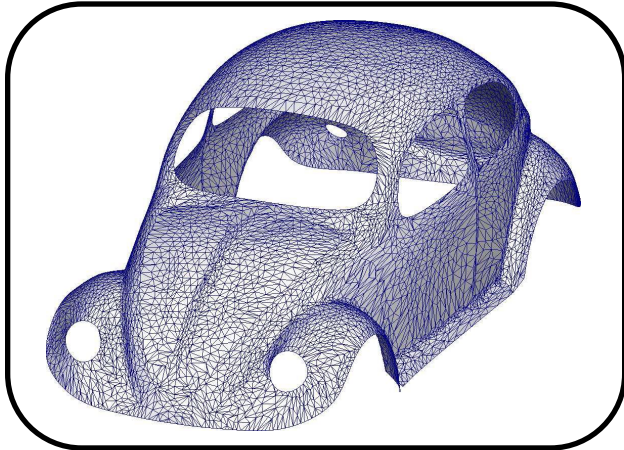
1 million vertices, 2 million triangles:  
Cannot be used in CAD/CAM software



# Zoom on Geometry Processing

## 1. Surface approximation – the challenge

output of the scanner



CAD/CAM software



Wrong **representation**,  
CAD/CAM needs **equations** instead of **samples**

**Q:** How can we "find the equation" of this car ?

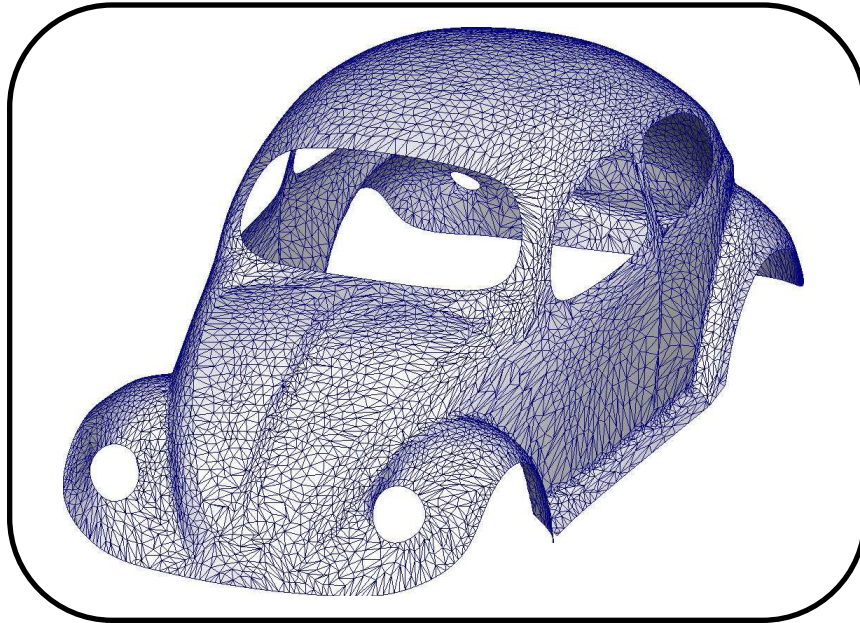




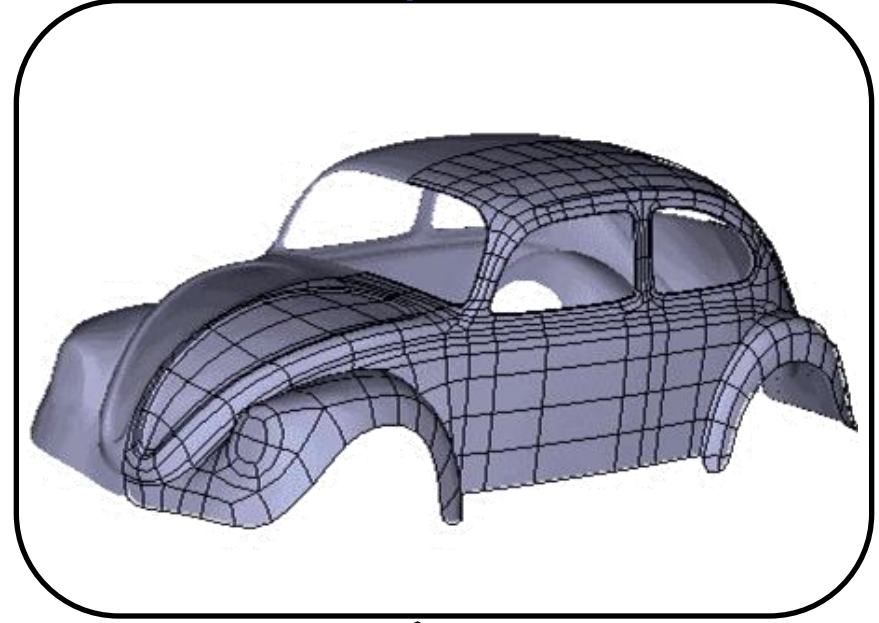
# Zoom on Geometry Processing

## 1. Surface approximation – the challenge

output of the scanner



CAD/CAM representation



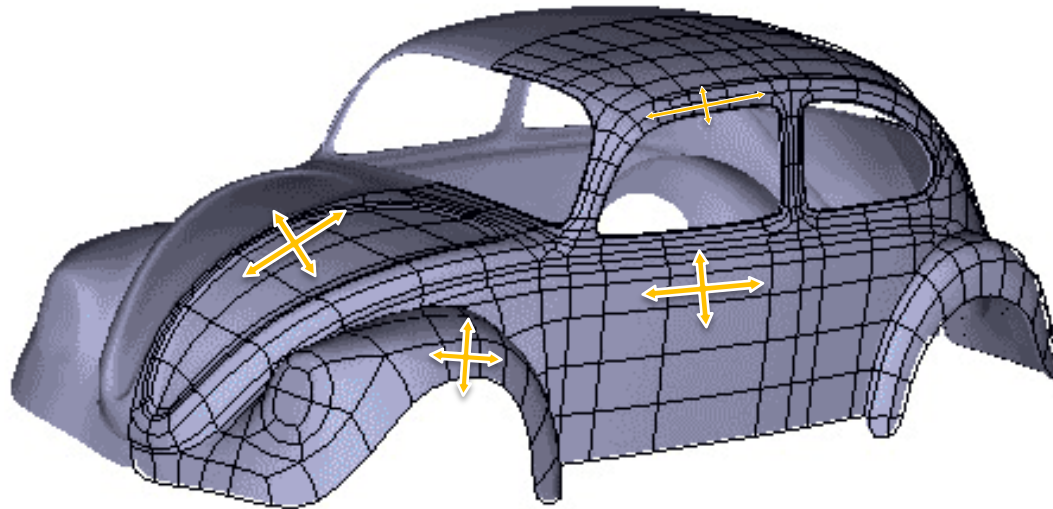
*1 million points → 100 Splines (cubic equations)*

**Q:** How can we "find the equation" of this car ?



# Zoom on Geometry Processing

## 2. Anisotropy and direction field design



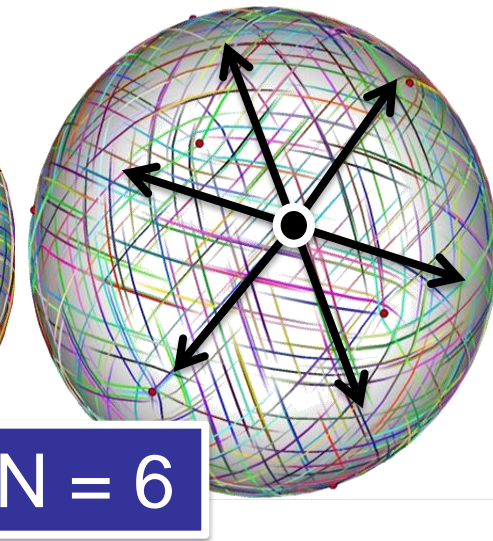
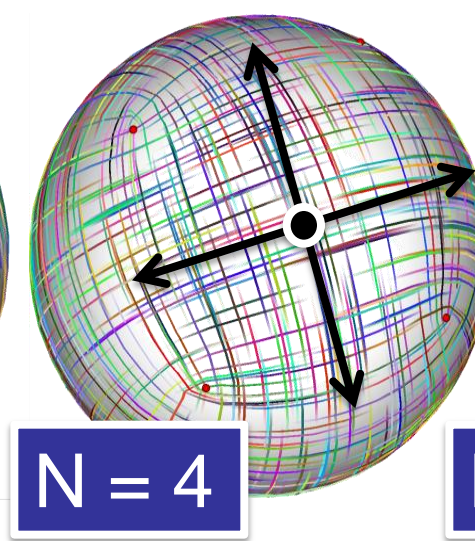
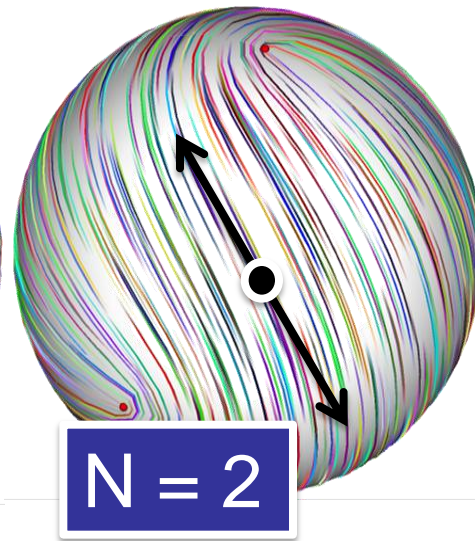
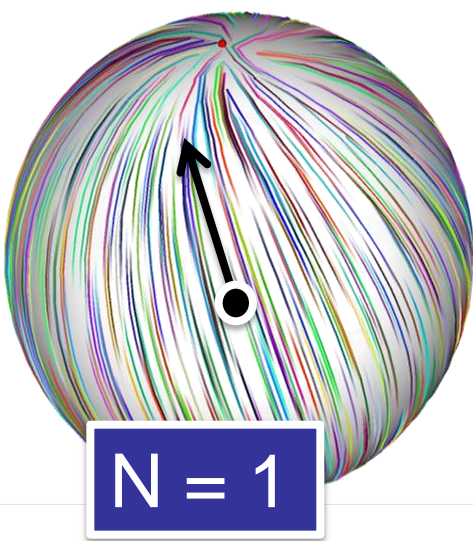
**Q:** How can we control the orientation/shape/size of the mesh/basis elements ?



# Zoom on Geometry Processing

## 2. Anisotropy and direction field design

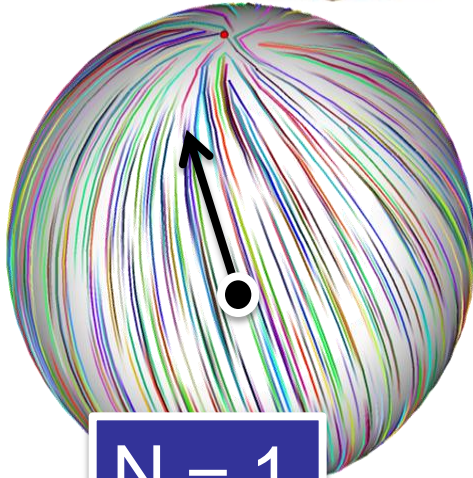
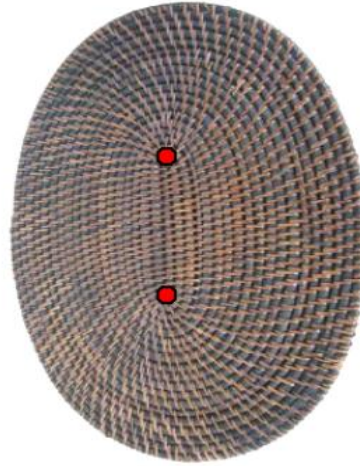
- A **N-symmetry direction field** is, for each point of a surface, a set of N unit vectors of the tangent plane that is invariant by rotation of  $2\pi/N$ .



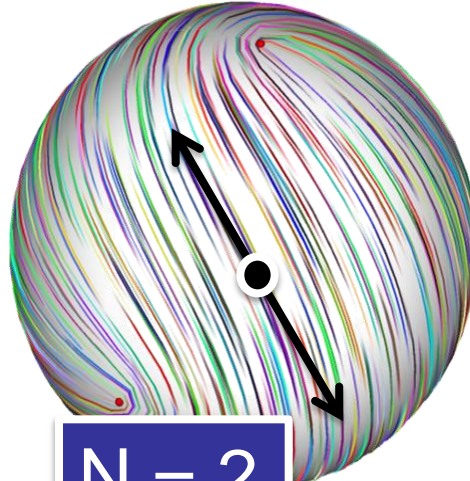


# Zoom on Geometry Processing

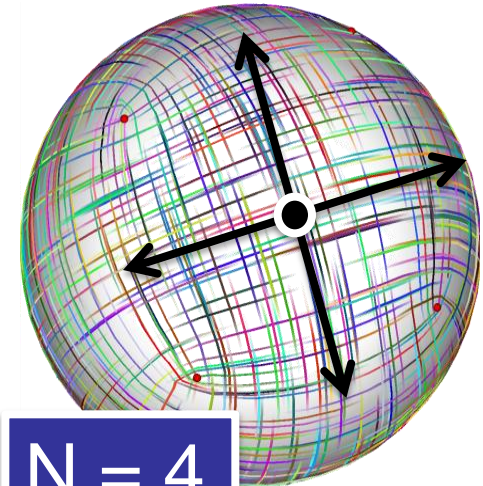
## 2. Anisotropy and direction field design



$N = 1$



$N = 2$



$N = 4$

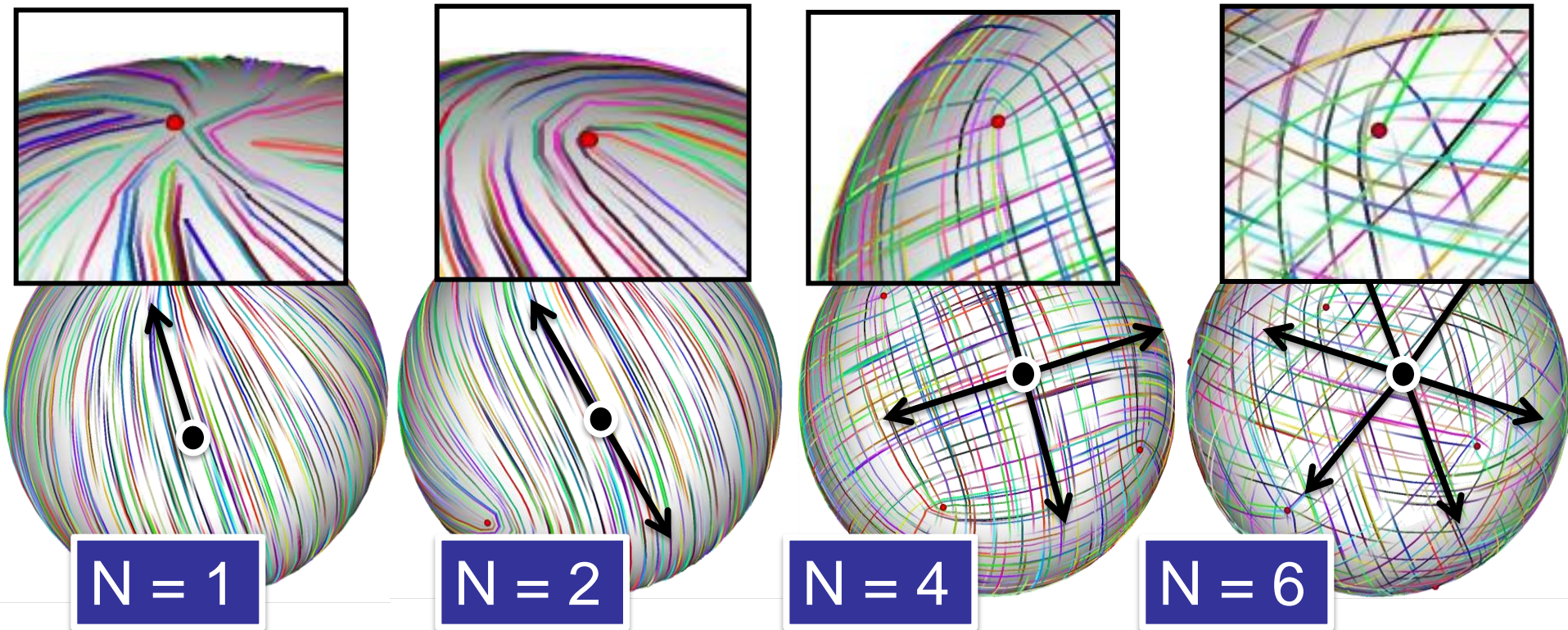




# Zoom on Geometry Processing

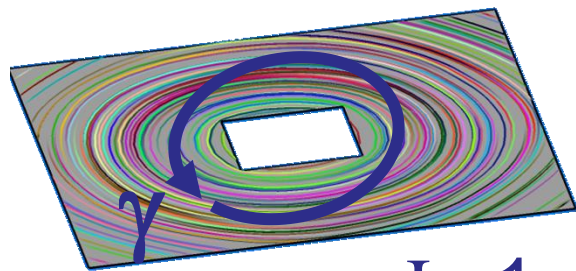
## 2. Anisotropy and direction field design

- **Singularities** generalize poles (and saddles) of vector fields.

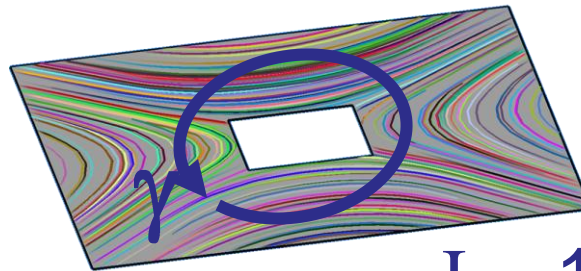


# Zoom on Geometry Processing

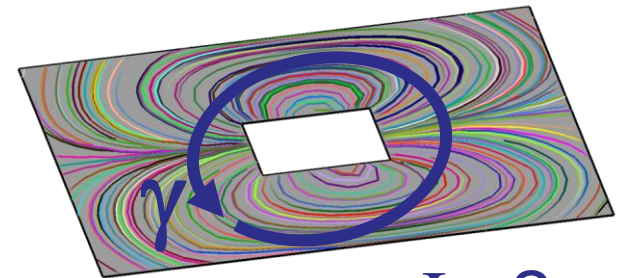
## 2. Anisotropy and direction field design



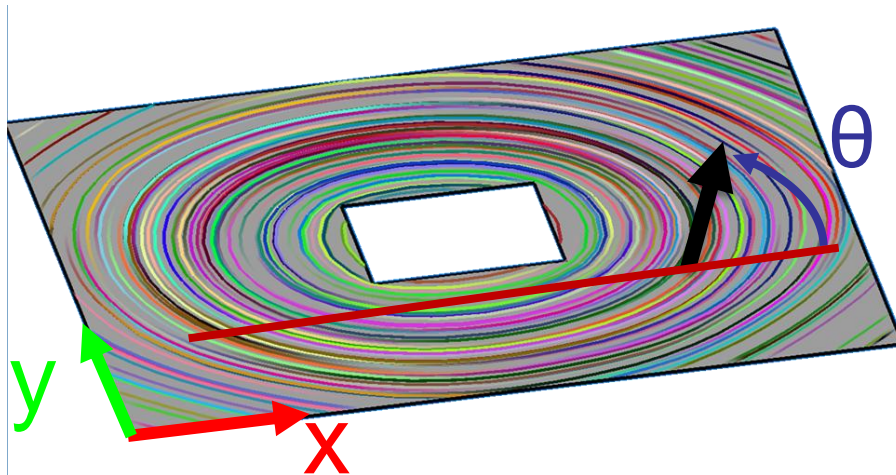
$N = 1$   $I = 1$



$N = 1$   $I = -1$



$N = 1$   $I = 2$



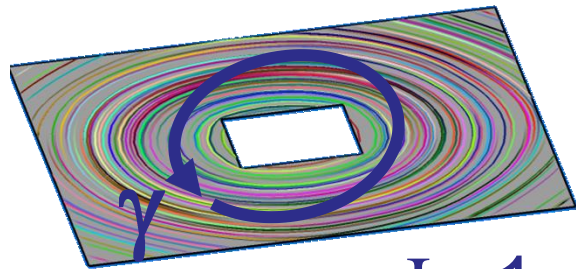
Index of a singularity

$$I = \int_{\gamma} d\theta / 2\pi$$

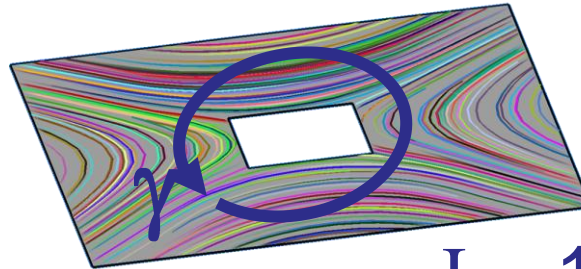


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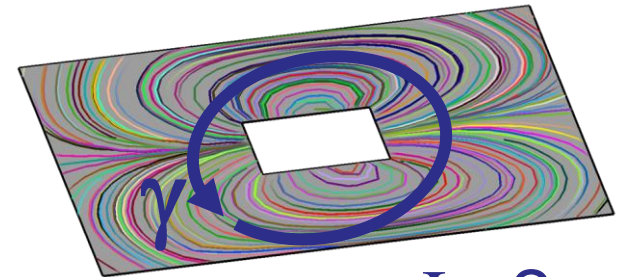
## 2. Anisotropy and direction field design



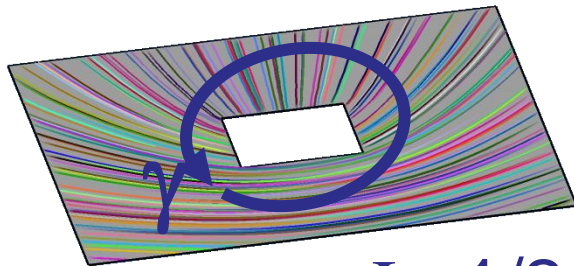
$N = 1$   $I = 1$



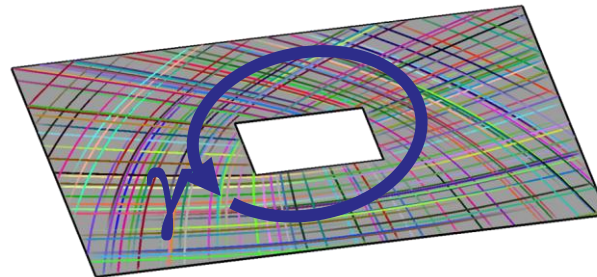
$N = 1$   $I = -1$



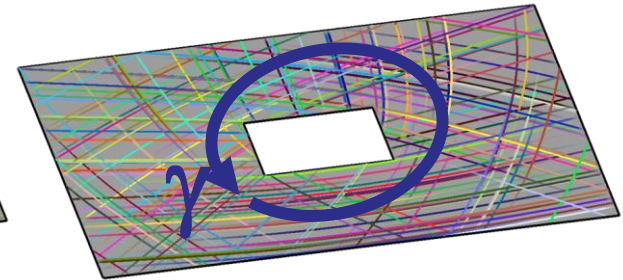
$N = 1$   $I = 2$



$N = 2$   $I = 1/2$



$N = 4$   $I = 1/4$

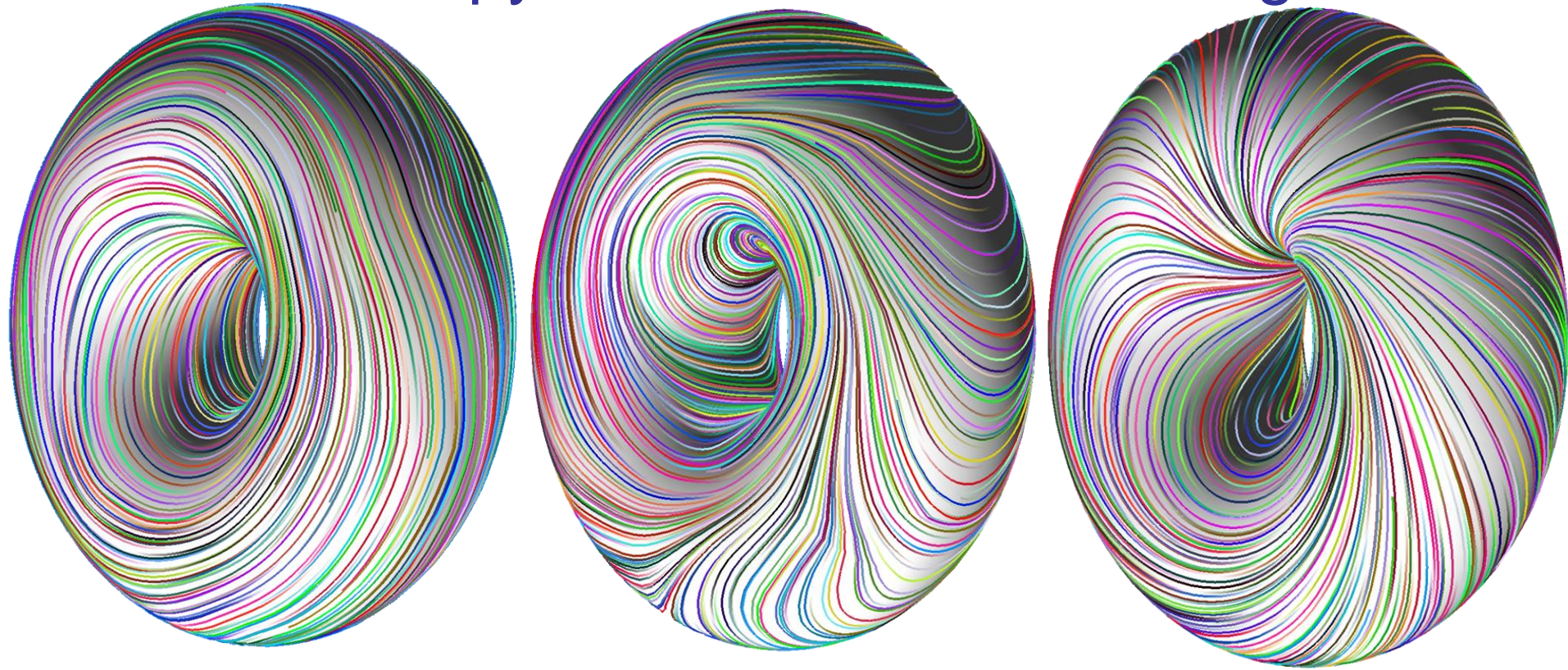


$N = 6$   $I = 1/6$



# Zoom on Geometry Processing

## 2. Anisotropy and direction field design



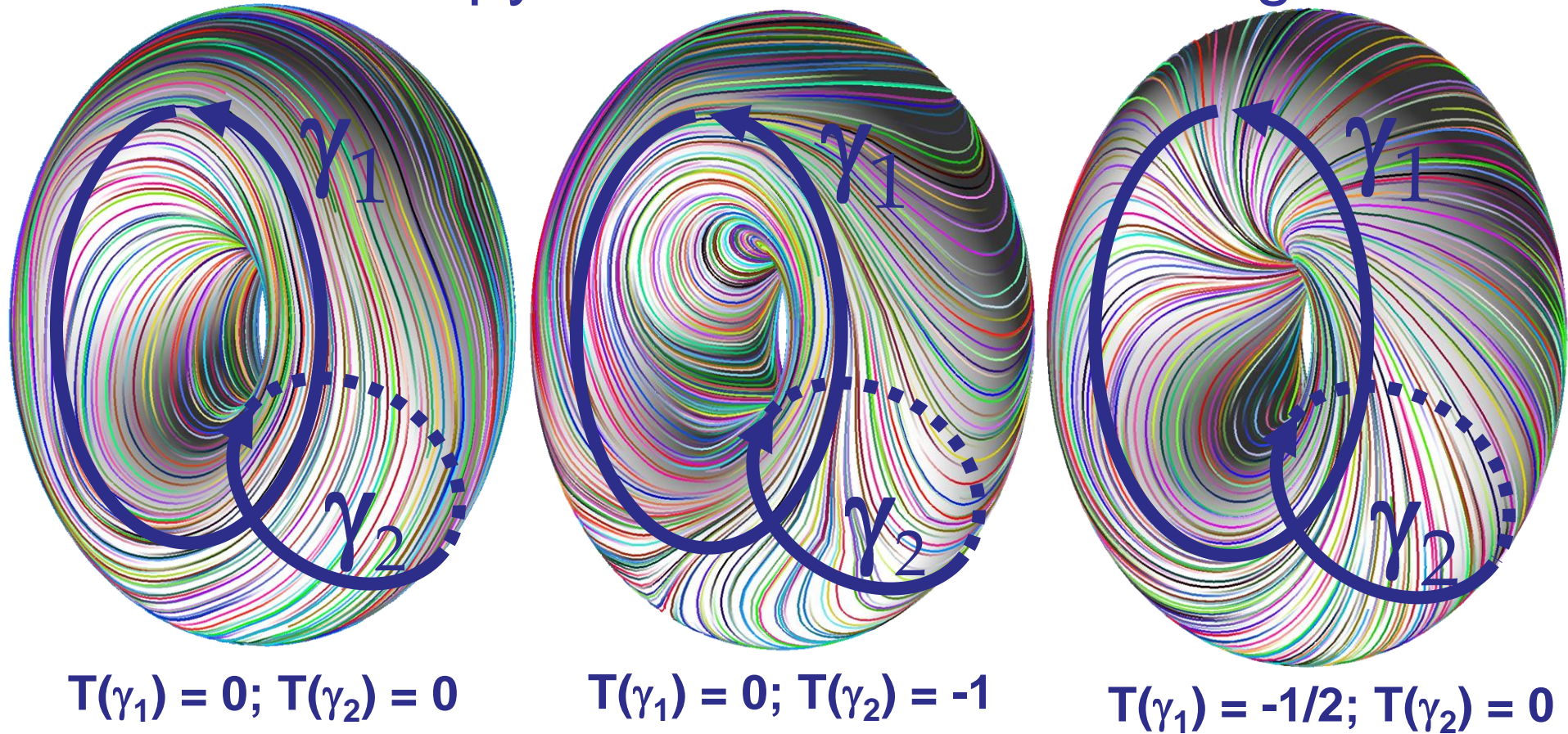
Arbitrary genus





# Zoom on Geometry Processing

## 2. Anisotropy and direction field design



Arbitrary genus  $\Rightarrow$  additional degrees of freedom



# Zoom on Geometry Processing

## 2. Anisotropy and direction field design

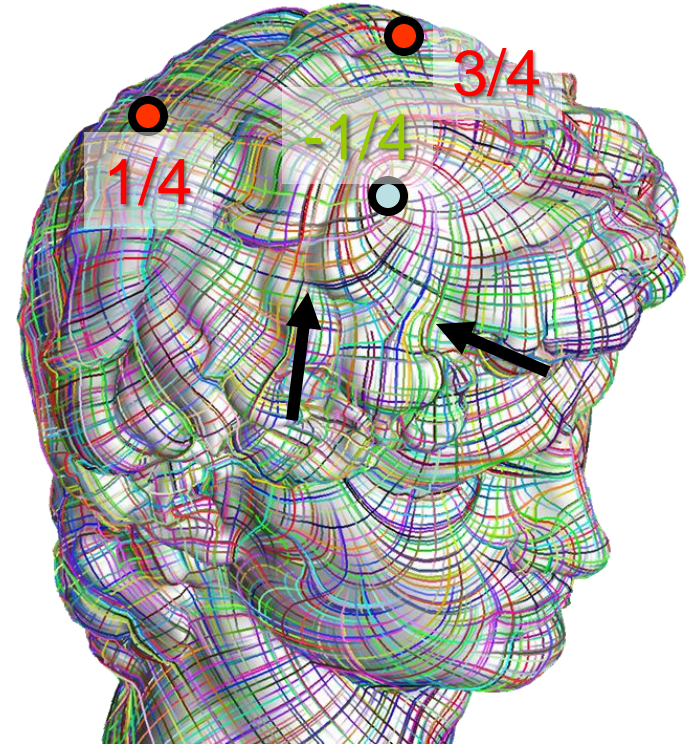
- Extension of the Poincaré-Hopf theorem to N-symmetry
- Discrete Index theory

[ACM TOG 2008]

$$\Sigma I = 2 - 2g$$

↑ Index                      ↑ Genus

Design with full topology control



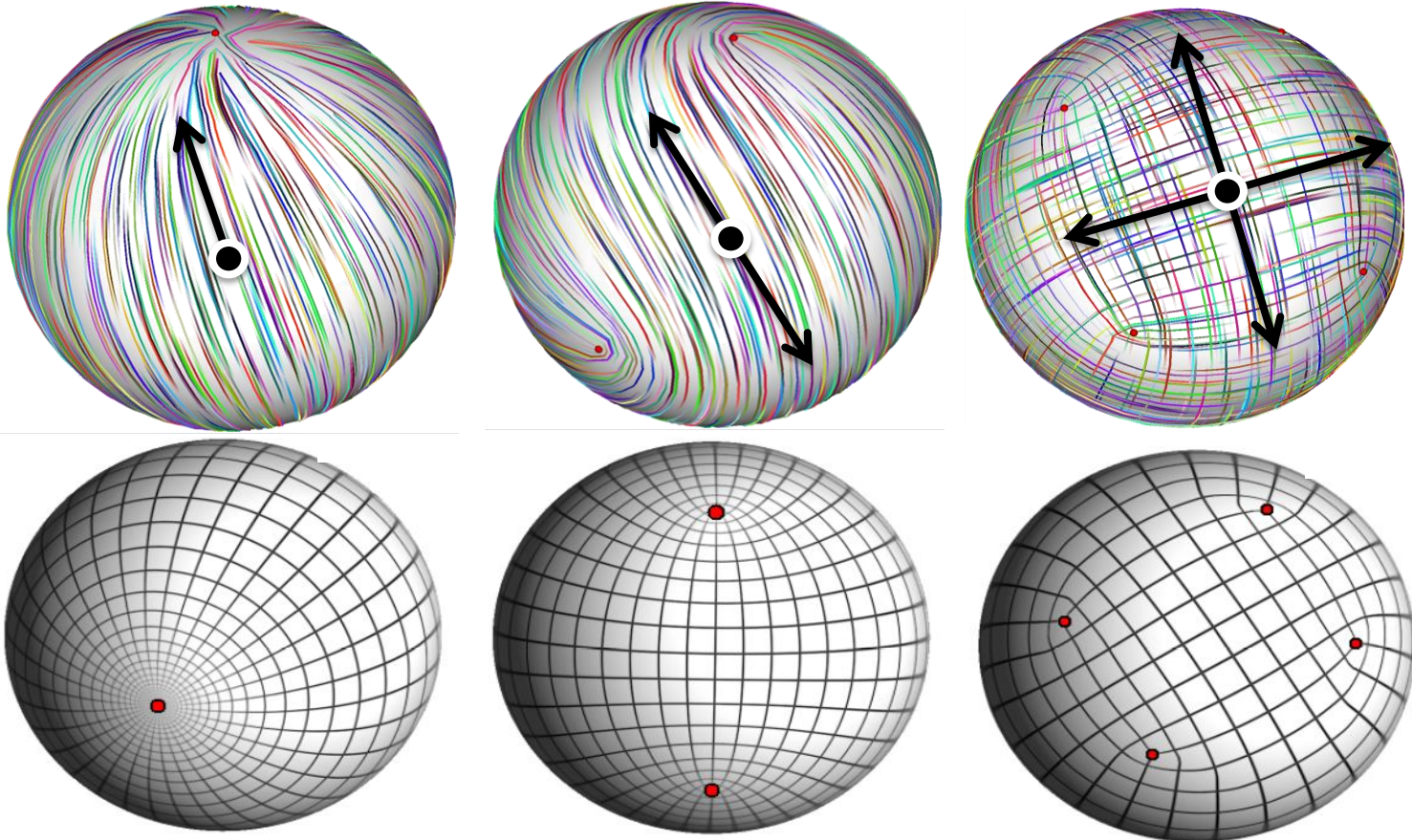
Controlled influence of geometry/topology [ACM TOG 2009]





# Zoom on Geometry Processing

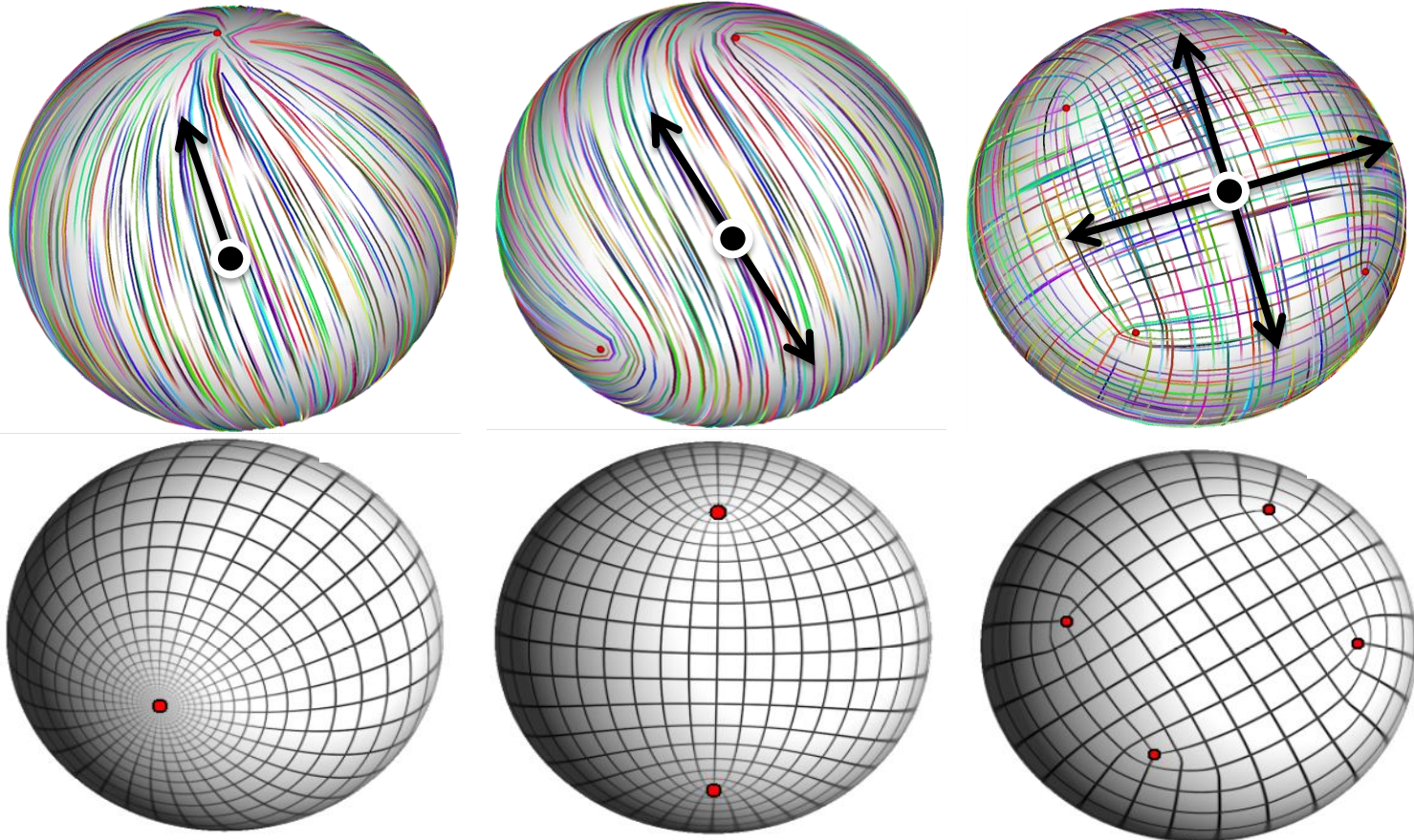
## 3. Global parameterization





# Zoom on Geometry Processing

## 3. Global parameterization

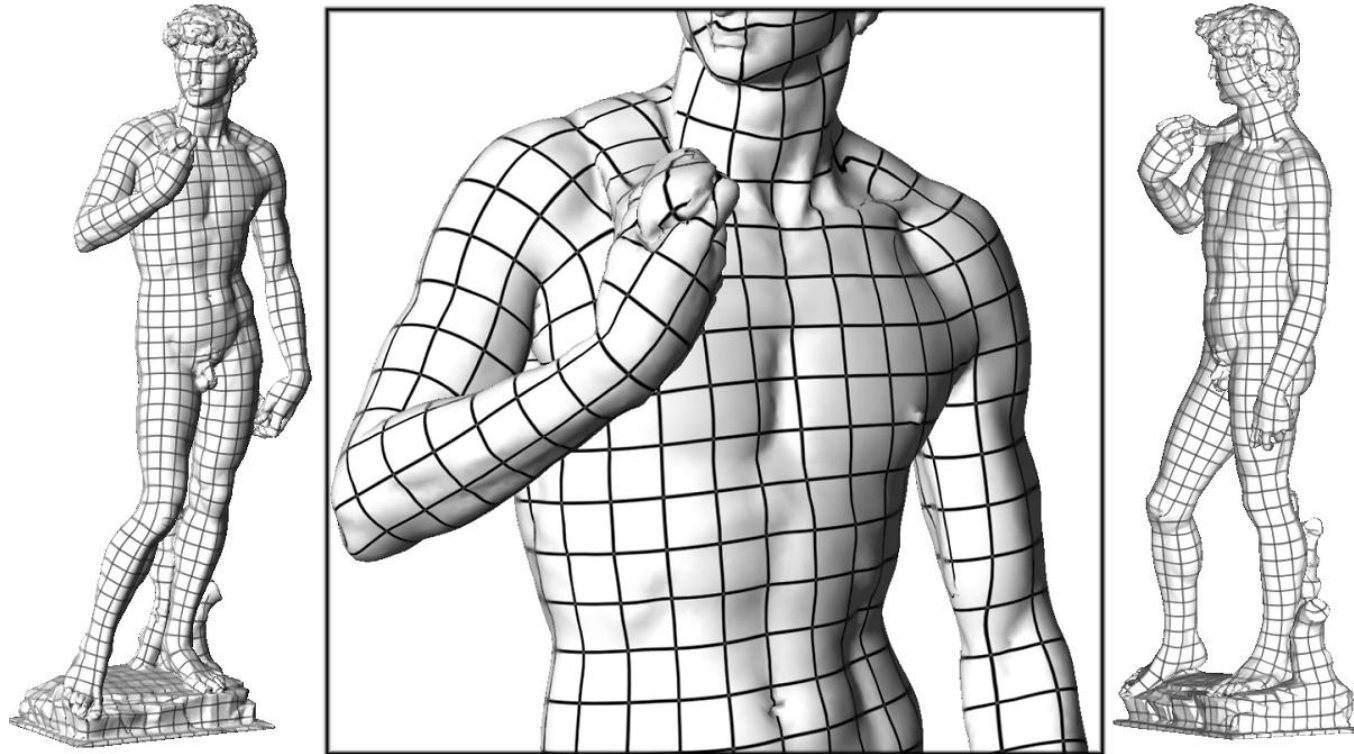


Q: How can we «integrate» a N-Symmetry direction field?



# Zoom on Geometry Processing

## 3. Global parameterization

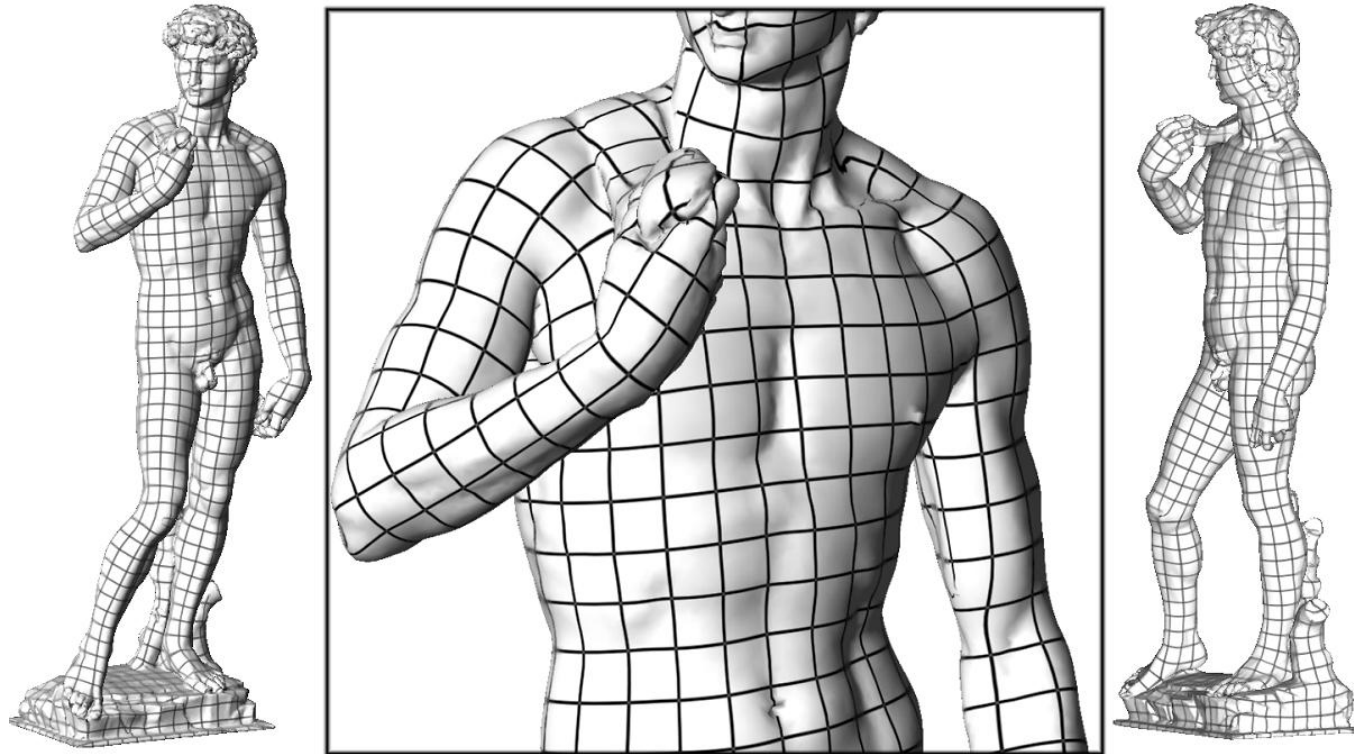


$$F^* = \sum_T \int_T \left( \|\nabla\theta^T - \omega\vec{K}_T\|^2 + \|\nabla\phi^T - \omega\vec{K}_T^\perp\|^2 \right) ds$$



# Zoom on Geometry Processing

## 3. Global parameterization



$$F_{T,i}^{\theta} \simeq \left\| U_{i\oplus 2} - \begin{pmatrix} \cos(\omega \vec{K}_i \cdot \vec{e}_i) & -\sin(\omega \vec{K}_i \cdot \vec{e}_i) \\ \sin(\omega \vec{K}_i \cdot \vec{e}_i) & \cos(\omega \vec{K}_i \cdot \vec{e}_i) \end{pmatrix} U_{i\oplus 1} \right\|^2$$

where:

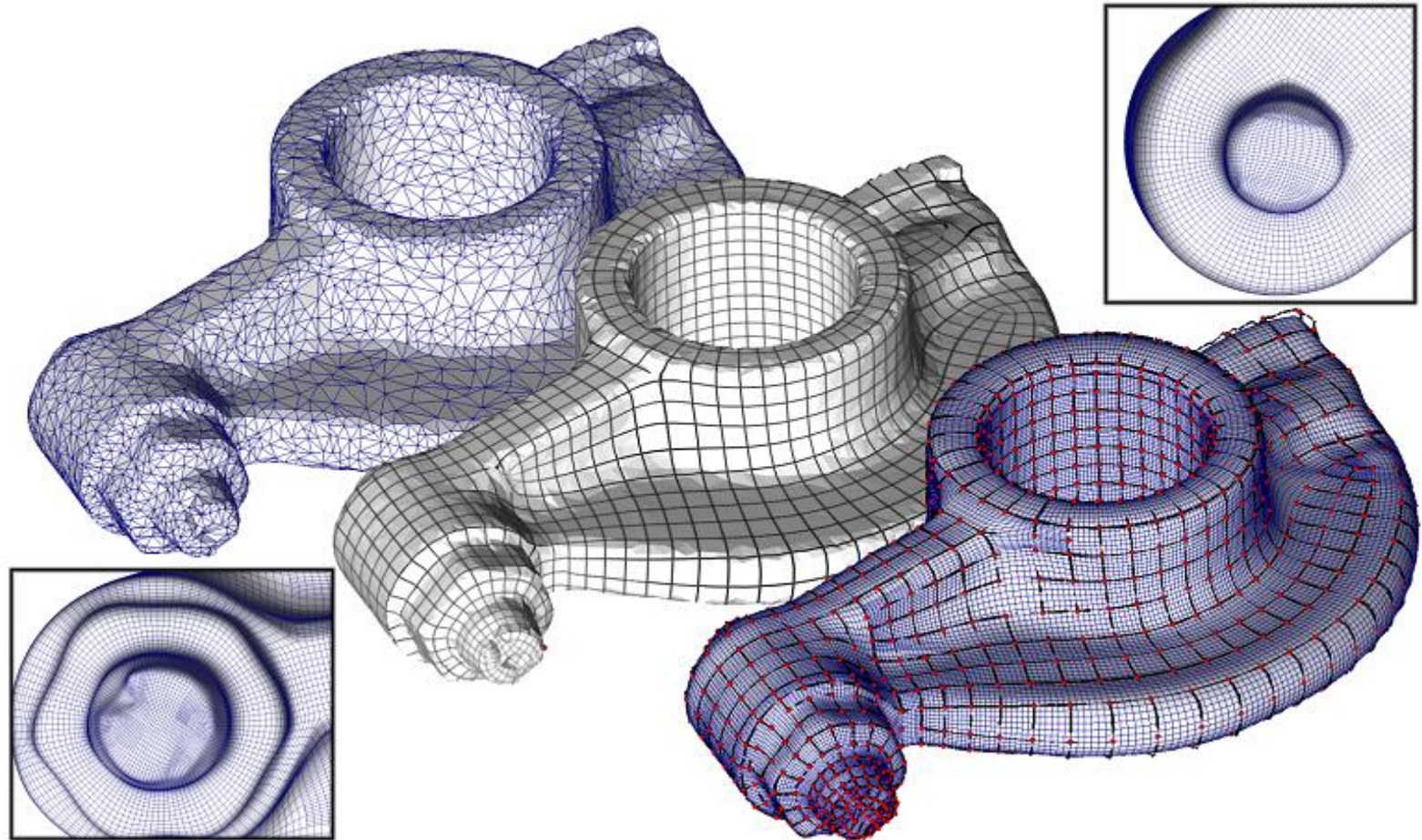
$$U_i = (\cos \theta_i, \sin \theta_i)$$





# Zoom on Geometry Processing

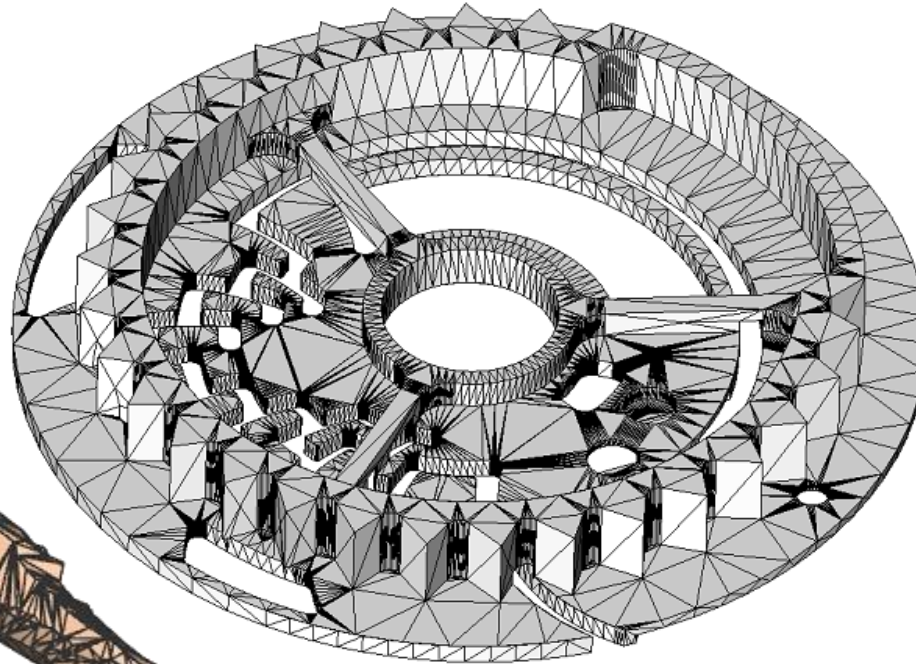
## 3. Global parameterization



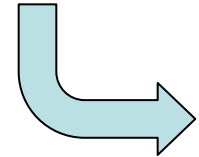


# Zoom on Geometry Processing

## 4. Optimal Sampling

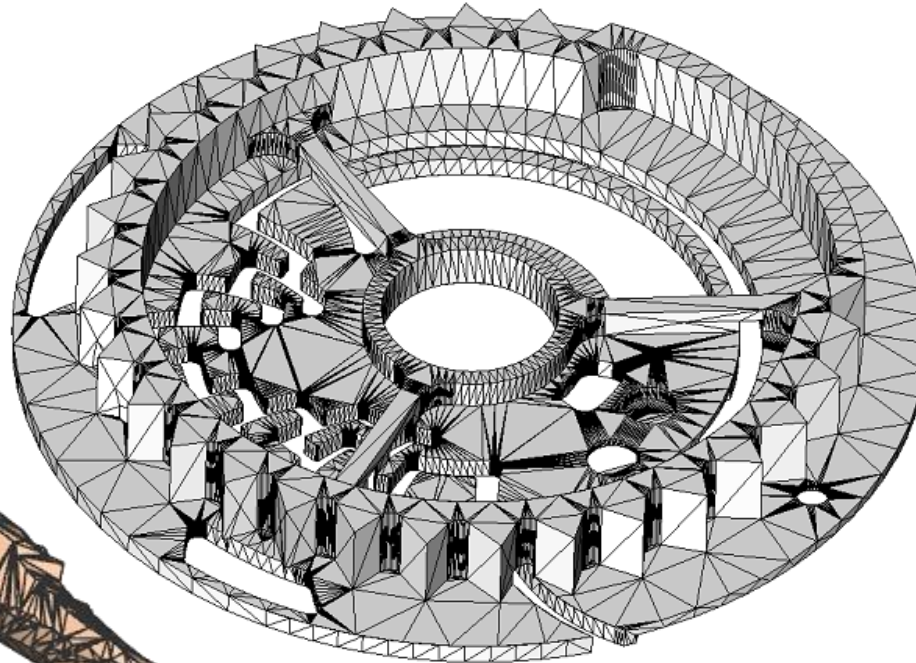


Lots of  
« needles »

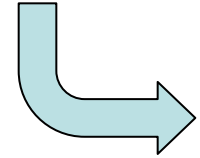


# Zoom on Geometry Processing

## 4. Optimal Sampling



Lots of  
« needles »



Q: How can we process these surfaces ?



# Zoom on Geometry Processing

## 4. Optimal Sampling



### Color quantization

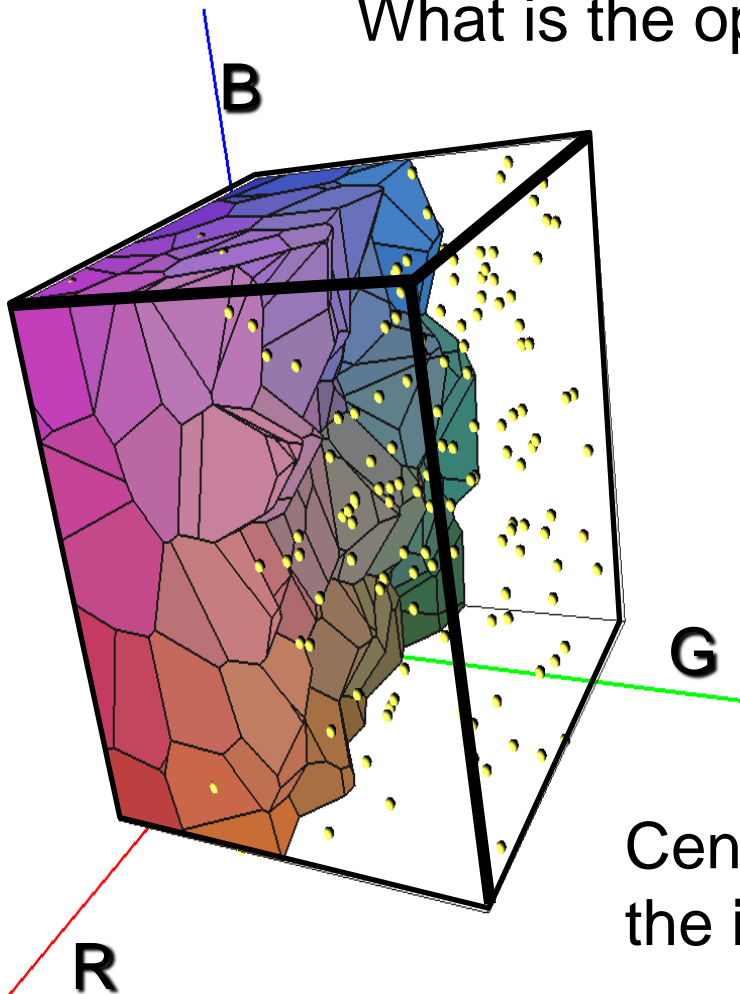
[Leung et.al, GPU Pro, AK Peters, 2010]



# Zoom on Geometry Processing

## 4. Optimal Sampling

What is the optimal colormap ?



Centroidal Voronoi Tessellation from the **information theory** perspective...

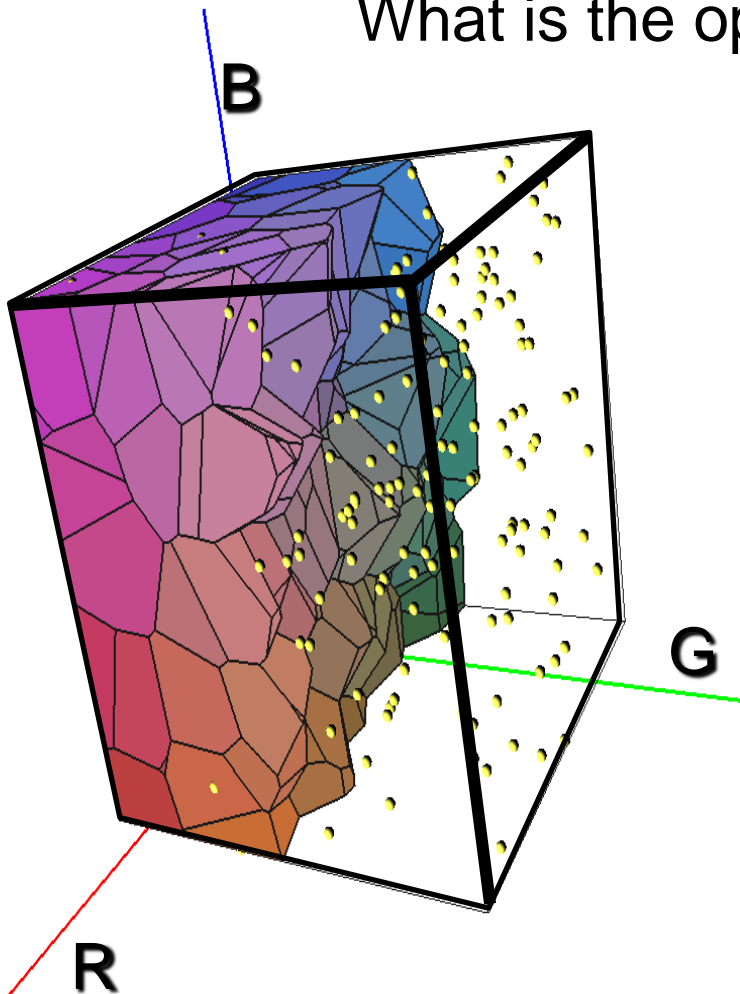




# Zoom on Geometry Processing

## 4. Optimal Sampling

What is the optimal colormap ?



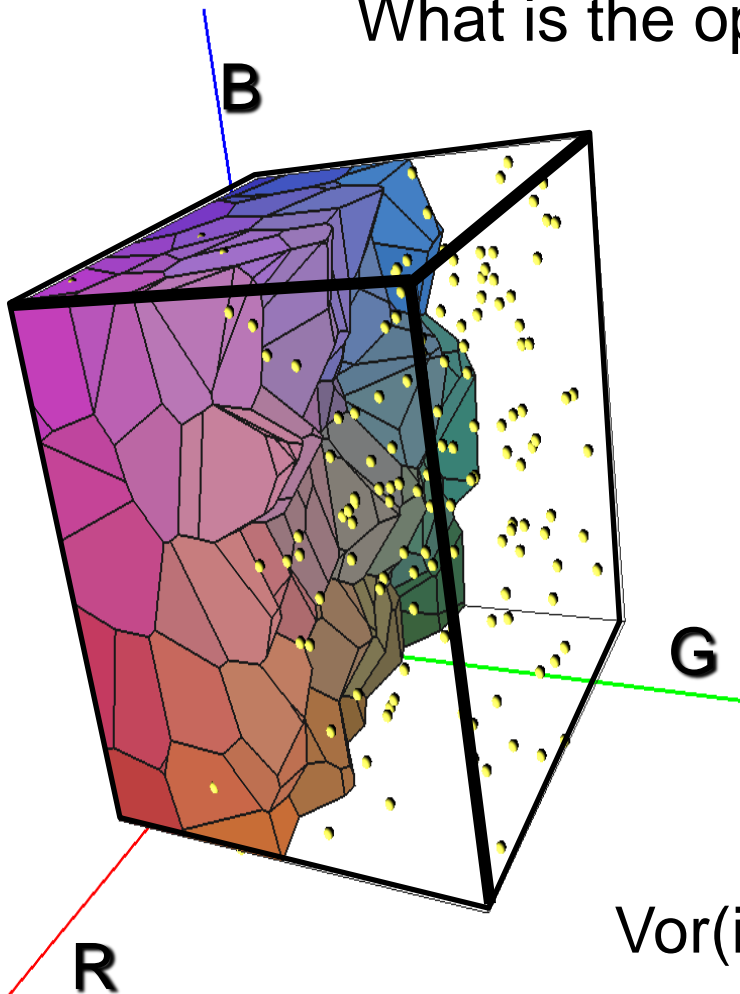
$\mathbf{x}_i = (r_i, g_i, b_i)$  Colormap entry



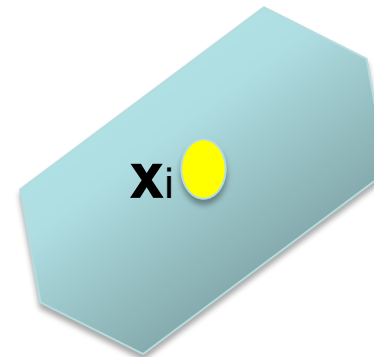
# Zoom on Geometry Processing

## 4. Optimal Sampling

What is the optimal colormap ?



$\mathbf{x}_i = (r_i, g_i, b_i)$  Colormap entry



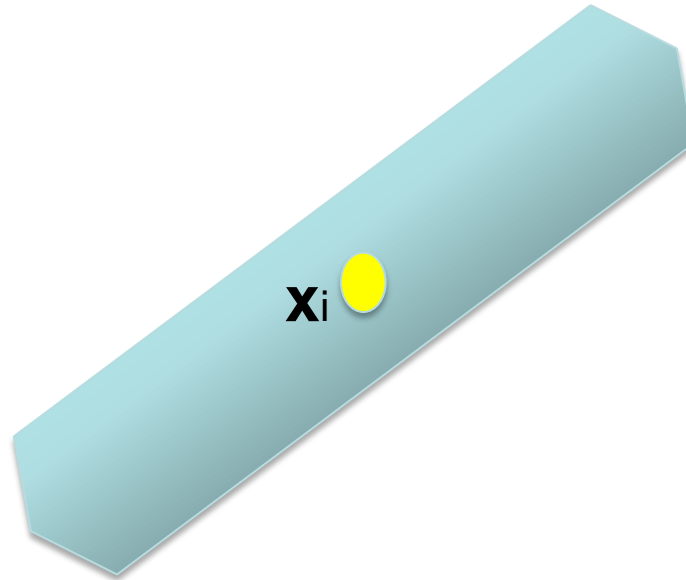
$$\text{Vor}(i) = \{ \mathbf{x} / d(\mathbf{x}, \mathbf{x}_i) < d(\mathbf{x}, \mathbf{x}_j) \} \forall i \neq j$$



# Zoom on Geometry Processing

## 4. Optimal Sampling

What is the optimal colormap ?



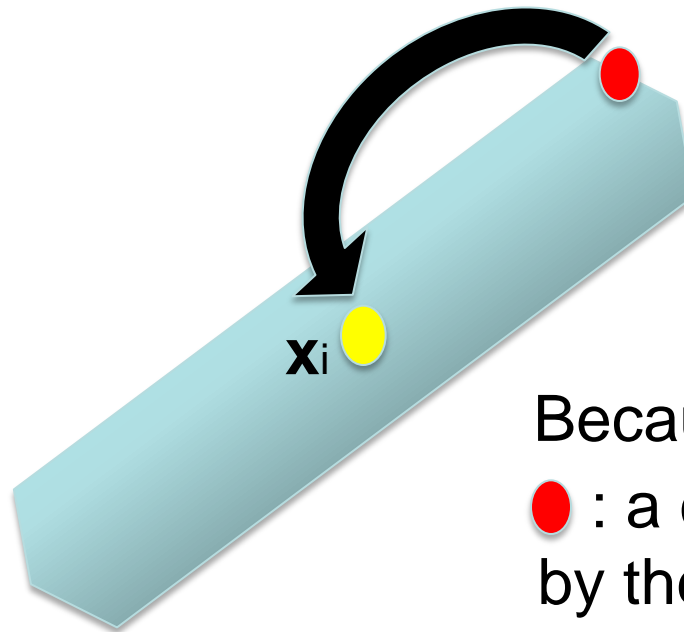
A « bad » colormap entry / Voronoi cell



# Zoom on Geometry Processing

## 4. Optimal Sampling

What is the optimal colormap ?



Why bad ?

Because  $\text{Vor}(x_i)$  contains  
● : a color poorly approximated  
by the colormap entry  $x_i$  ●

A « bad » colormap entry / Voronoi cell

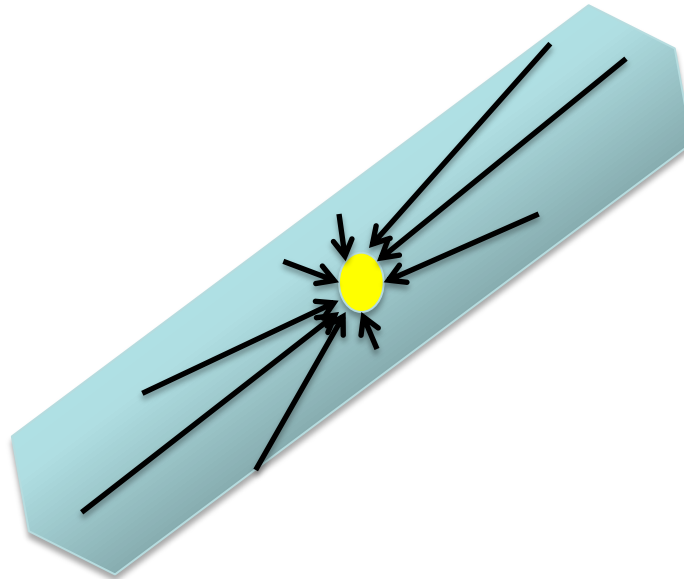




# Zoom on Geometry Processing

## 4. Optimal Sampling

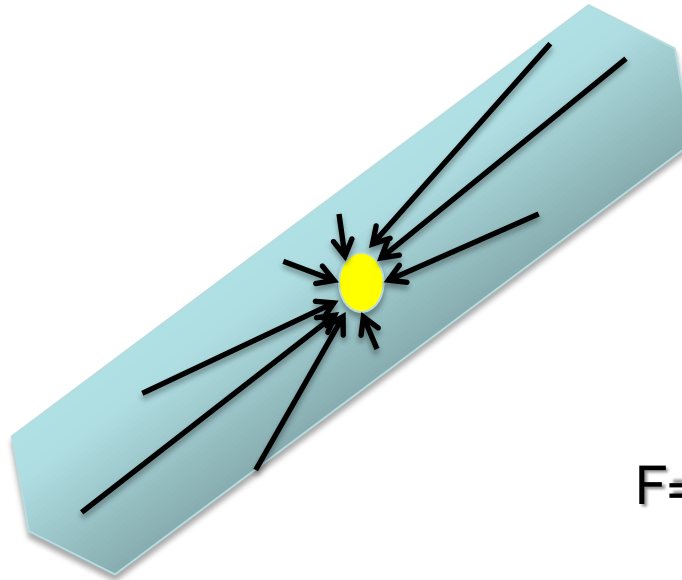
What is the optimal colormap ?



# Zoom on Geometry Processing

## 4. Optimal Sampling

What is the optimal colormap ?



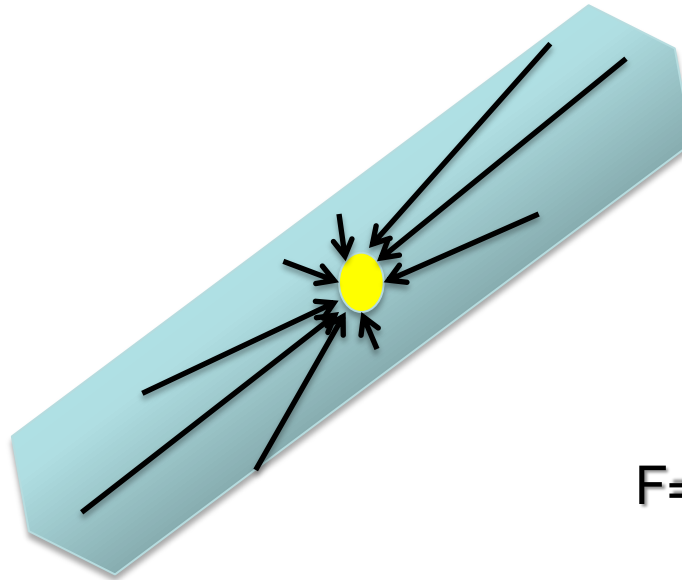
$$F = \int_{\text{Vor}(i)} \left\| \mathbf{x}_i - \mathbf{x} \right\|^2 d\mathbf{x}$$



# Zoom on Geometry Processing

## 4. Optimal Sampling

What is the optimal colormap ?



$$F = \int_{\text{Vor}(i)} \left\| \mathbf{x}_i - \mathbf{x} \right\|^2 d\mathbf{x}$$

**F: Quantization noise power**

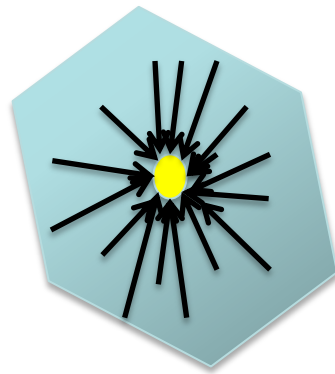




# Zoom on Geometry Processing

## 4. Optimal Sampling

What is the optimal colormap ?



$$F = \int_{\text{Vor}(i)} \left\| \mathbf{x}_i - \mathbf{x} \right\|^2 d\mathbf{x}$$

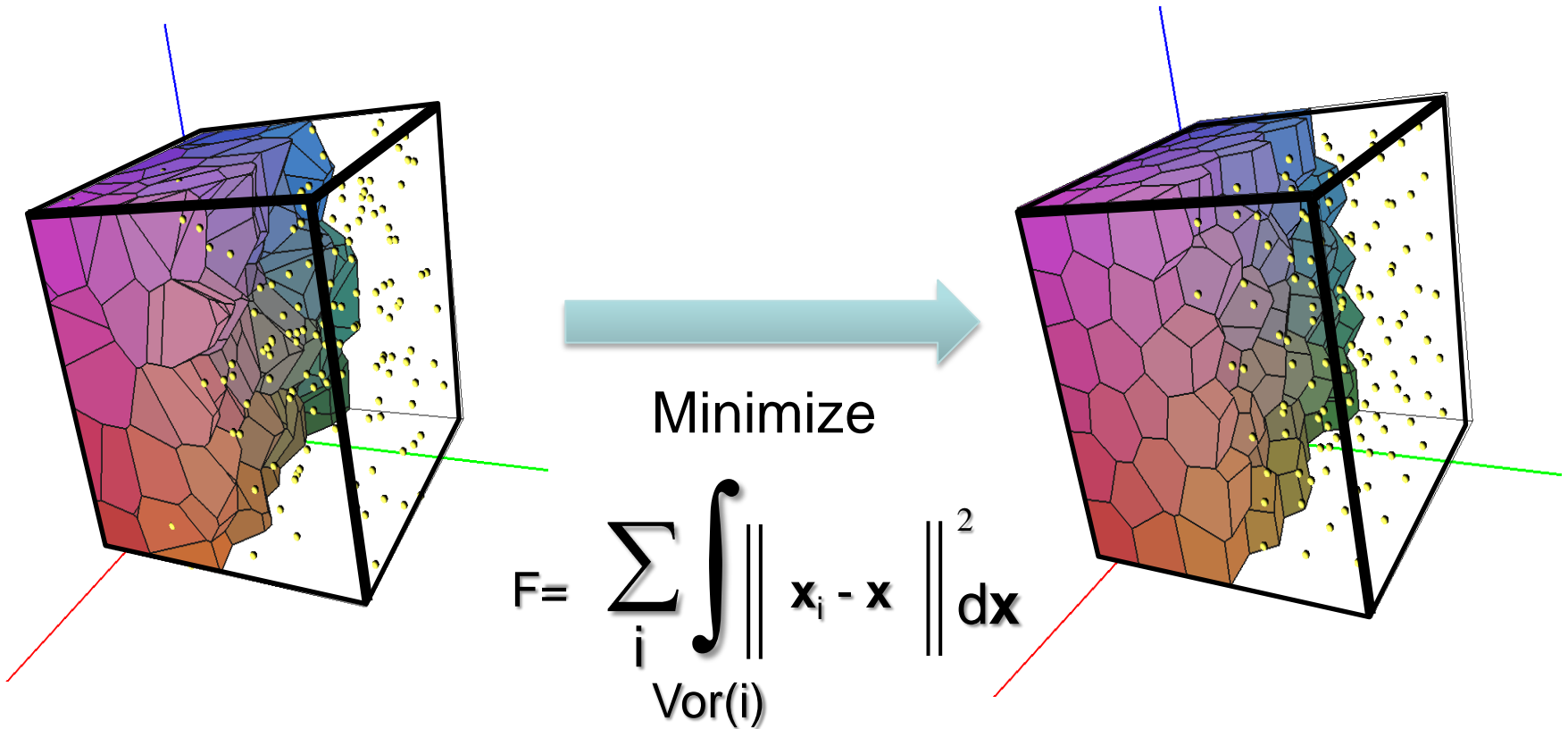
**F: Quantization noise power**



# Zoom on Geometry Processing

## 4. Optimal Sampling

What is the optimal colormap ?



# Zoom on Geometry Processing

## 4. Optimal Sampling

The classical method:

Lloyd's algorithm = gradient descent

$$F = \sum_i \int_{\text{Vor}(i)} \left\| \mathbf{x}_i - \mathbf{x} \right\|^2 d\mathbf{x}$$





# Zoom on Geometry Processing

## 4. Optimal Sampling

### Lloyd's Relaxation:

(Geometric point of view)

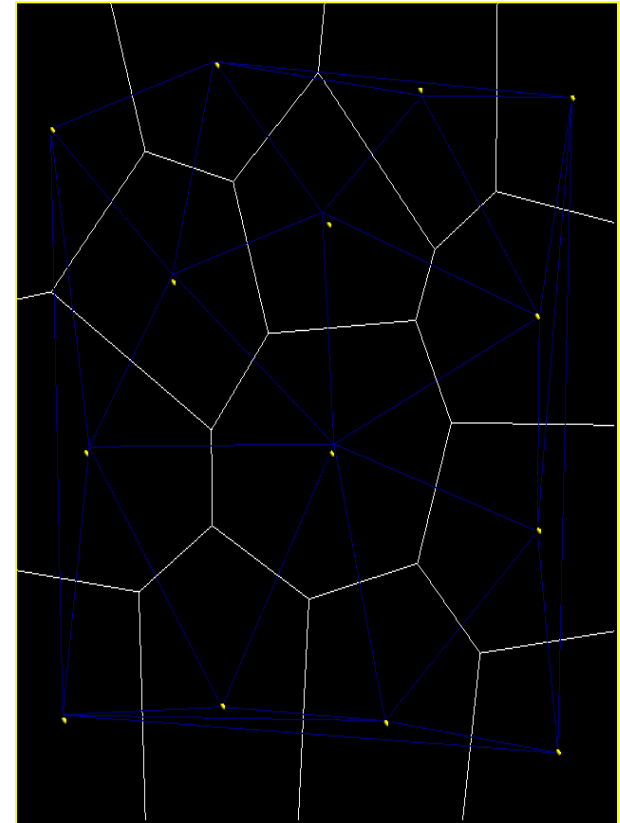
Loop

Move the  $\mathbf{x}_i$ 's to the  $\mathbf{g}_i$ 's

Re-triangulate

End loop

- + Provably decreases  $F$  [Du et.al]
- + Reasonably easy to implement
- Slow (linear) convergence



# Zoom on Geometry Processing

## 4. Optimal Sampling

[ACM TOG 2009] (information theory point of view)

Newton's method for minimizing multivariate non-linear function  $F$

While  $|\nabla F| > \varepsilon$

solve

$$\begin{bmatrix} \nabla^2_{x,x} F \end{bmatrix} \begin{bmatrix} \delta X \end{bmatrix} = - \begin{bmatrix} \nabla_x F \end{bmatrix}$$

$$X \leftarrow X + \delta X$$

End while

Hessian = 2<sup>nd</sup> order derivatives  
Is  $F$  sufficiently continuous? ( $C^2$ )

Yes [Liu, Wang, L, Sun, Yan, Lu and Yang 09]



# Zoom on Geometry Processing

## 4. Optimal Sampling

*CVT in 2D*



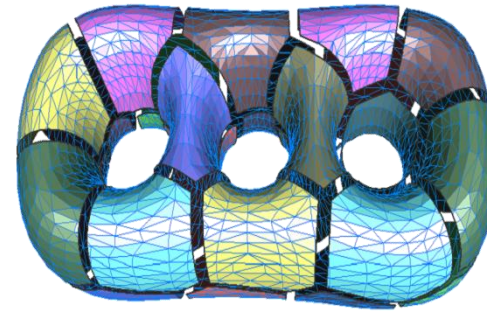


# Zoom on Geometry Processing

## 4. Optimal Sampling

*CVT in 2D*

*CVT on surfaces*



[Yan, L, Liu, Sun and Wang SGP2009]



# Zoom on Geometry Processing

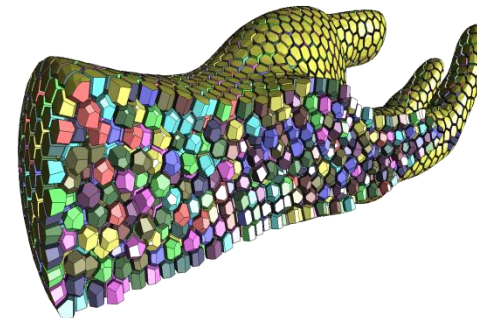
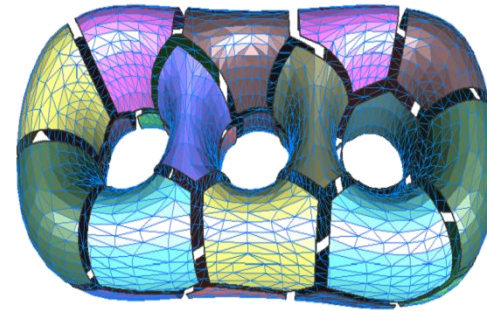
## 4. Optimal Sampling

*CVT in 2D*

*CVT on surfaces*

*CVT in volumes*

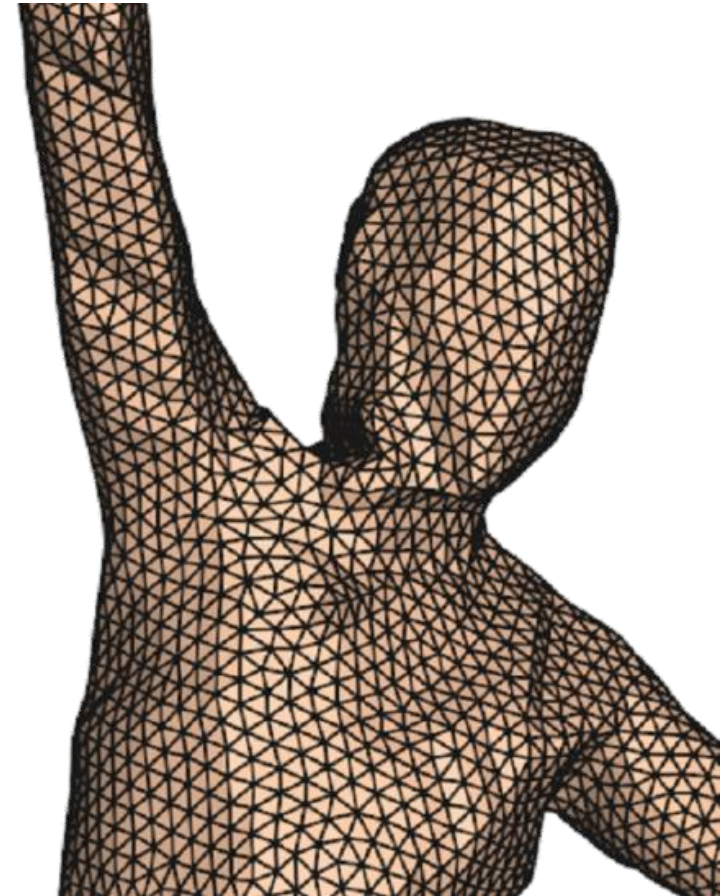
[Yan, Wang, L, Liu 2010]



# Zoom on Geometry Processing

## 4. Optimal Sampling

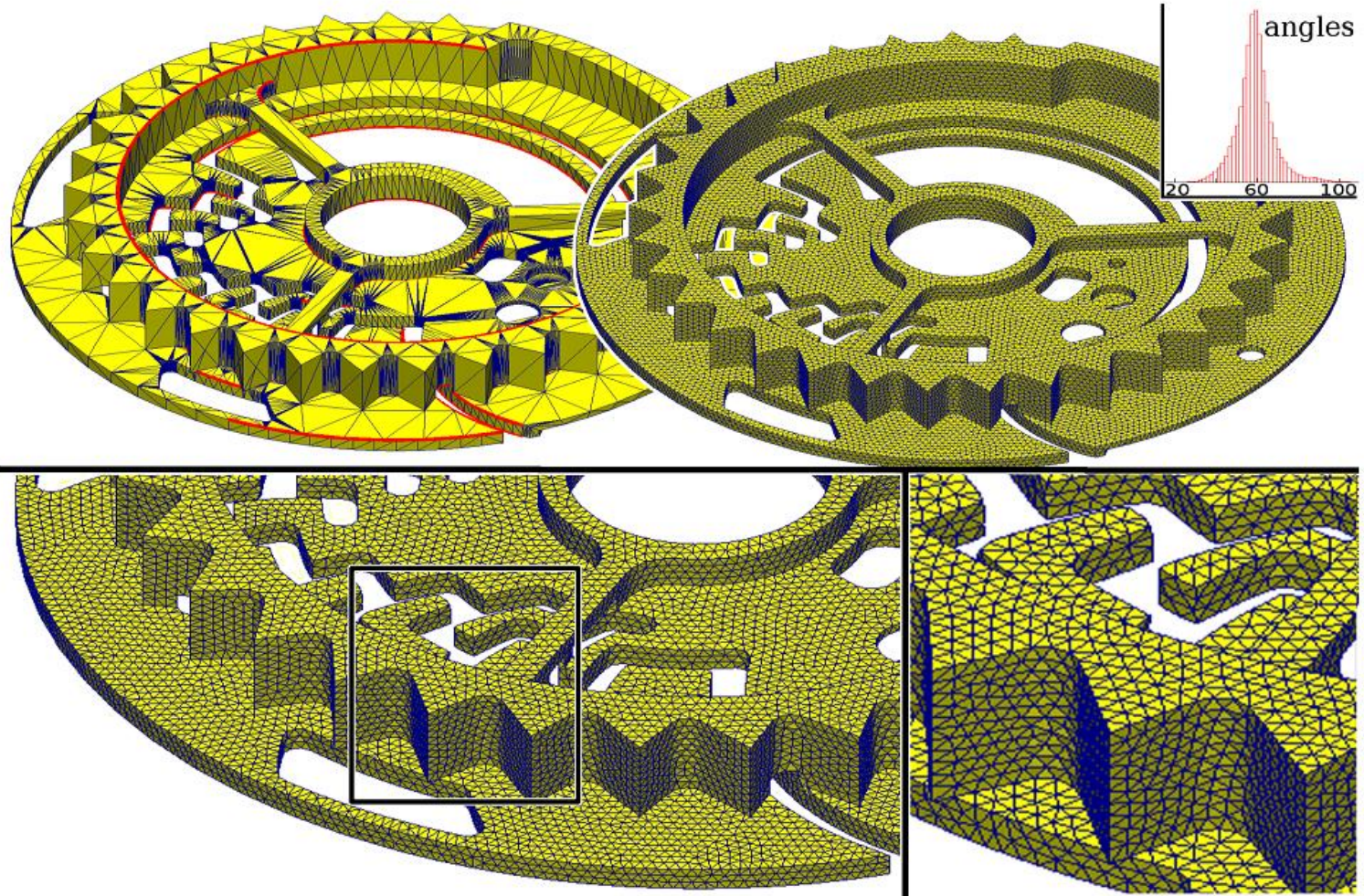
Remeshing [Yan, L, Liu, Sun and Wang – SGP2009]





# Zoom on Geometry Processing

## 4. Optimal Sampling





# Zoom on Geometry Processing

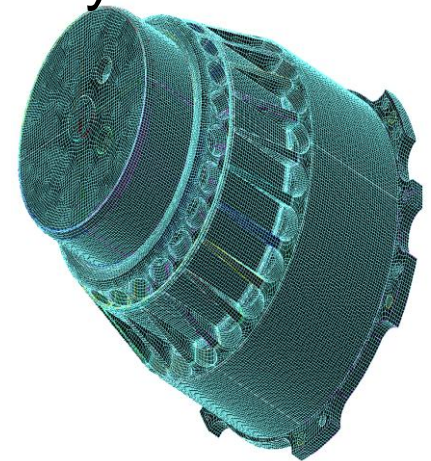
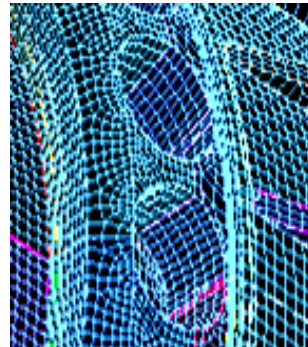
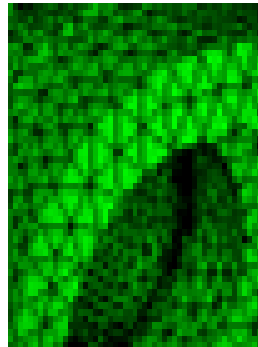
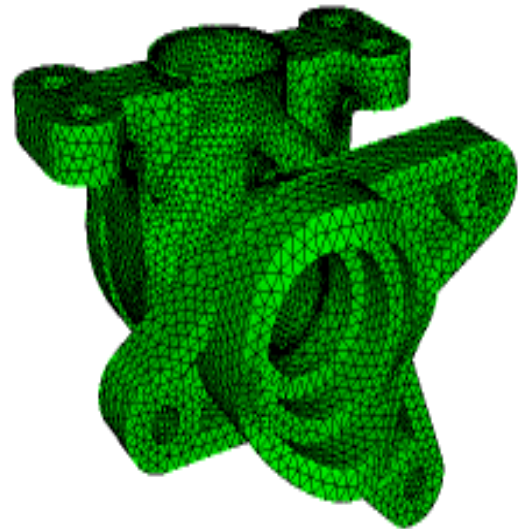
## 4. Optimal Sampling

### ***Tet Meshing***

1. Fully Automated
2. Millions of elements in minutes/seconds
3. Adequate for some analysis
4. Inaccurate for other Analysis

### ***Hex Meshing***

1. Partially Automated, some Manual
2. Millions of elements in days/weeks/months
3. Preferred by some analysts for solution quality

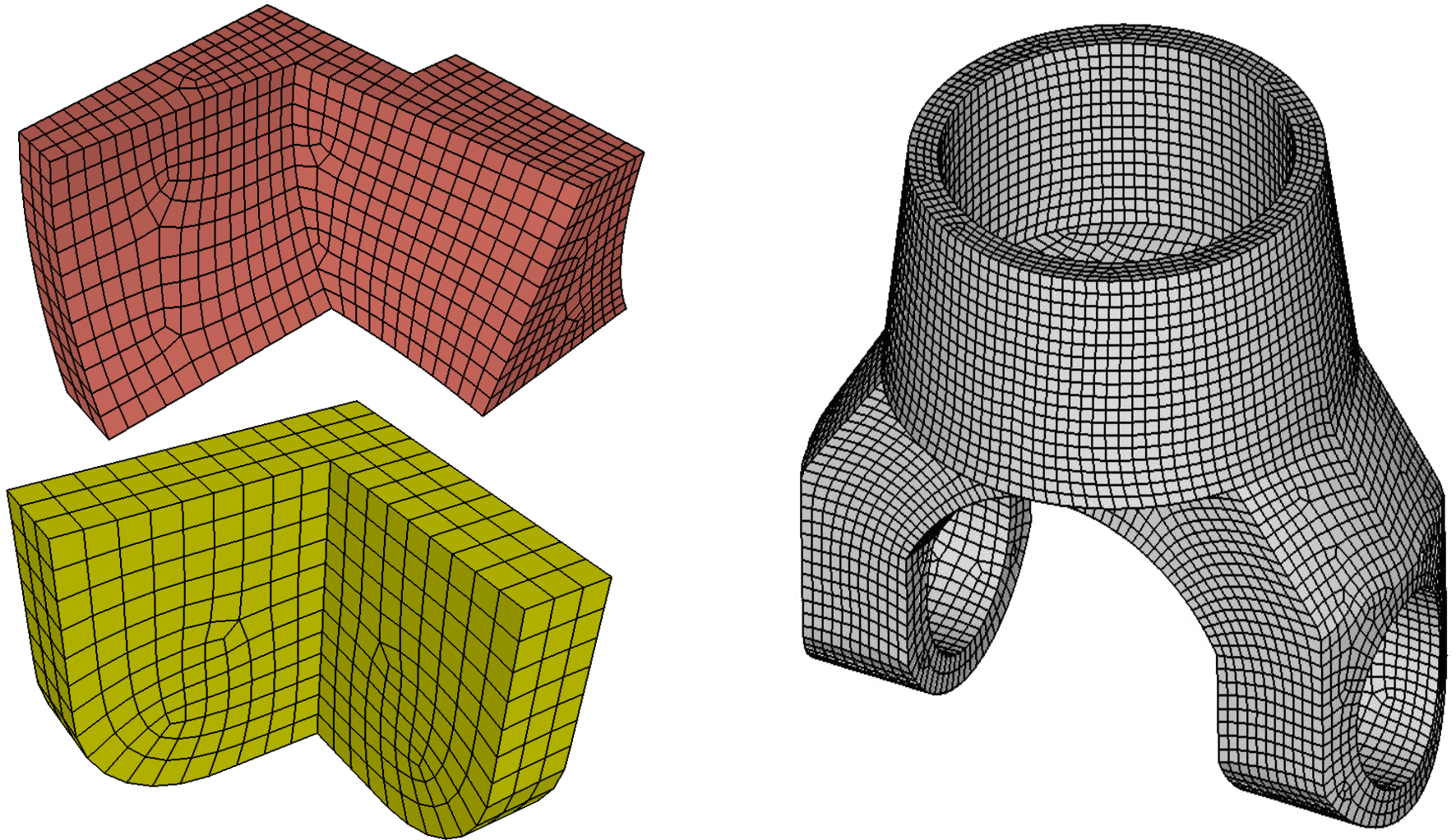


[Matt Staten] (Sandial Labs)



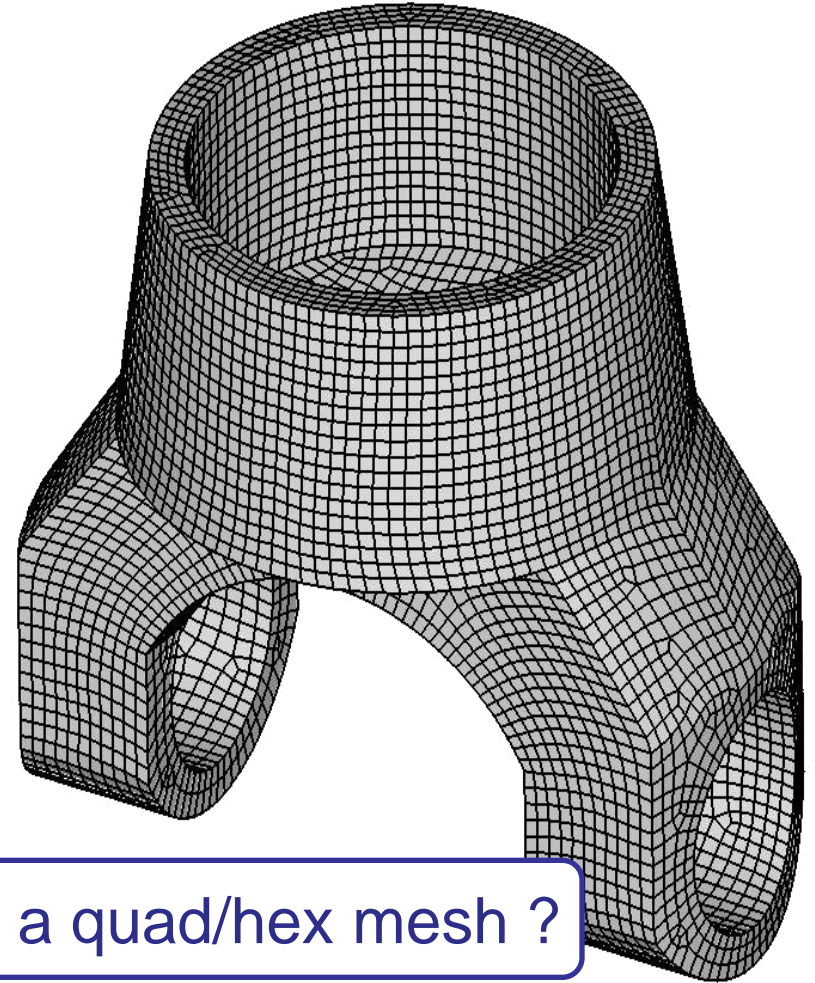
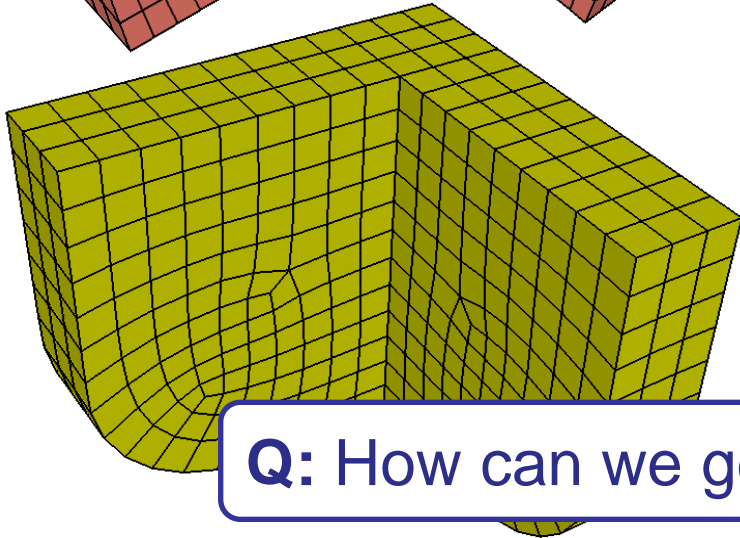
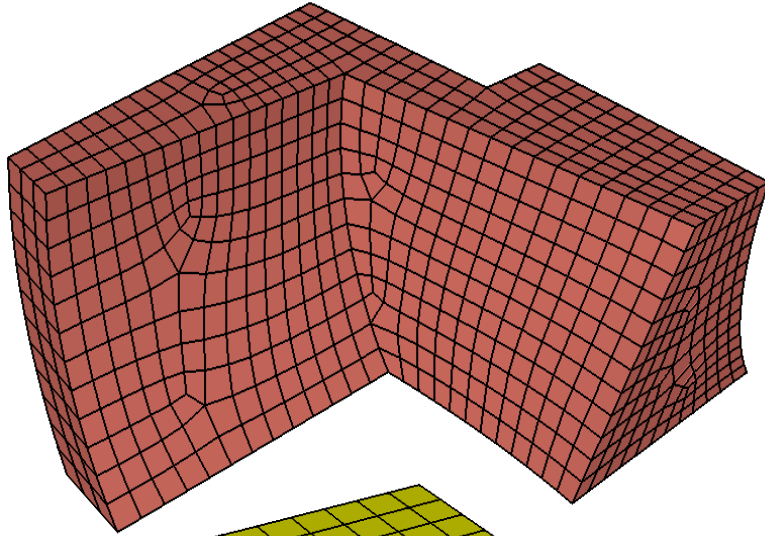
# Zoom on Geometry Processing

## 4. Optimal Sampling



# Zoom on Geometry Processing

## 4. Optimal Sampling



Q: How can we generate a quad/hex mesh ?



# Zoom on Geometry Processing

## 4. Optimal Sampling

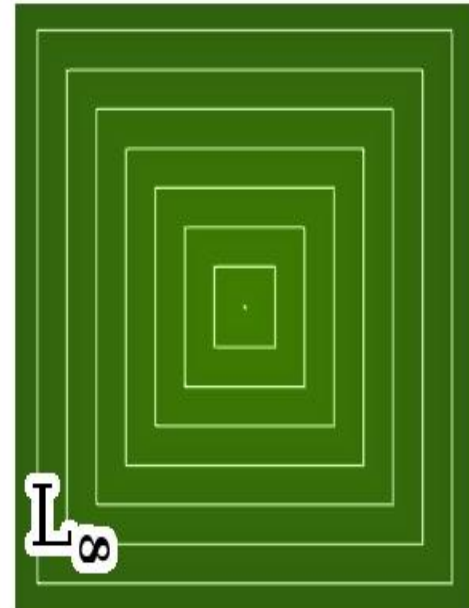
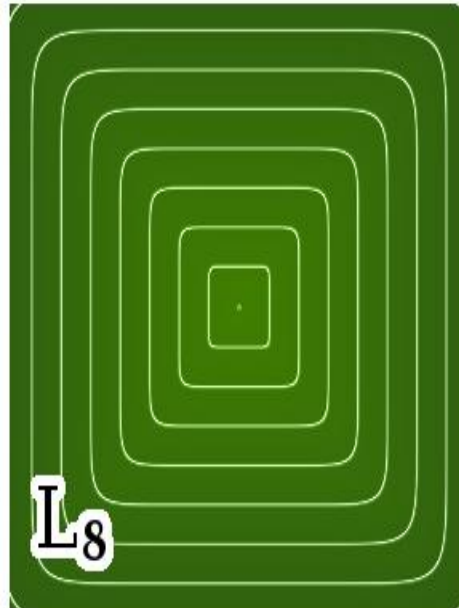
### Blowing Square Bubbles ...

$p=2$

$p=4$

$p=8$

....





# Zoom on Geometry Processing

## 4. Optimal Sampling

Standard CVT: 
$$F = \sum_i \int_{\text{Vor}(i)} \left\| \mathbf{x}_i - \mathbf{x} \right\|^2 d\mathbf{x}$$



# Zoom on Geometry Processing

## 4. Optimal Sampling

Standard CVT:  $F = \sum_i \int_{\text{Vor}(i)} \left\| \mathbf{x}_i - \mathbf{x} \right\|^2 d\mathbf{x}$

Lp CVT:  
[L and Liu 2010]  $F = \sum_i \int_{\text{Vor}(i)} \left\| \mathbf{M}(\mathbf{x}) (\mathbf{x}_i - \mathbf{x}) \right\|_p^p d\mathbf{x}$



# Zoom on Geometry Processing

## 4. Optimal Sampling

L<sub>p</sub> CVT:

$$F = \sum_i \int_{\text{Vor}(i)} \left\| \mathbf{M}(\mathbf{x}) (\mathbf{x}_i - \mathbf{x}) \right\|_p^p d\mathbf{x}$$

Anisotropy, encodes desired orientation  
Riemannian metric  $\mathbf{G} = \mathbf{M}^t \mathbf{M}$



# Zoom on Geometry Processing

## 4. Optimal Sampling

L<sub>p</sub> CVT:

$$F = \sum_i \int_{\text{Vor}(i)} \left\| M(\mathbf{x}) (\mathbf{x}_i - \mathbf{x}) \right\|_p^p d\mathbf{x}$$

L<sub>p</sub> norm:  $\| \mathbf{x} \|_p = \sqrt[p]{|\mathbf{x}|^p + |\mathbf{y}|^p + |\mathbf{z}|^p}$

If  $p$  is even:  $\| \mathbf{x} \|_p^p = x^p + y^p + z^p$





# Zoom on Geometry Processing

## 4. Optimal Sampling

L<sub>p</sub> CVT: 
$$F = \sum_i \int_{\text{Vor}(i)} \left\| \mathbf{M}(\mathbf{x}) (\mathbf{x}_i - \mathbf{x}) \right\|_p^p d\mathbf{x}$$

### Optimization with LBFGS (quasi-Newton)

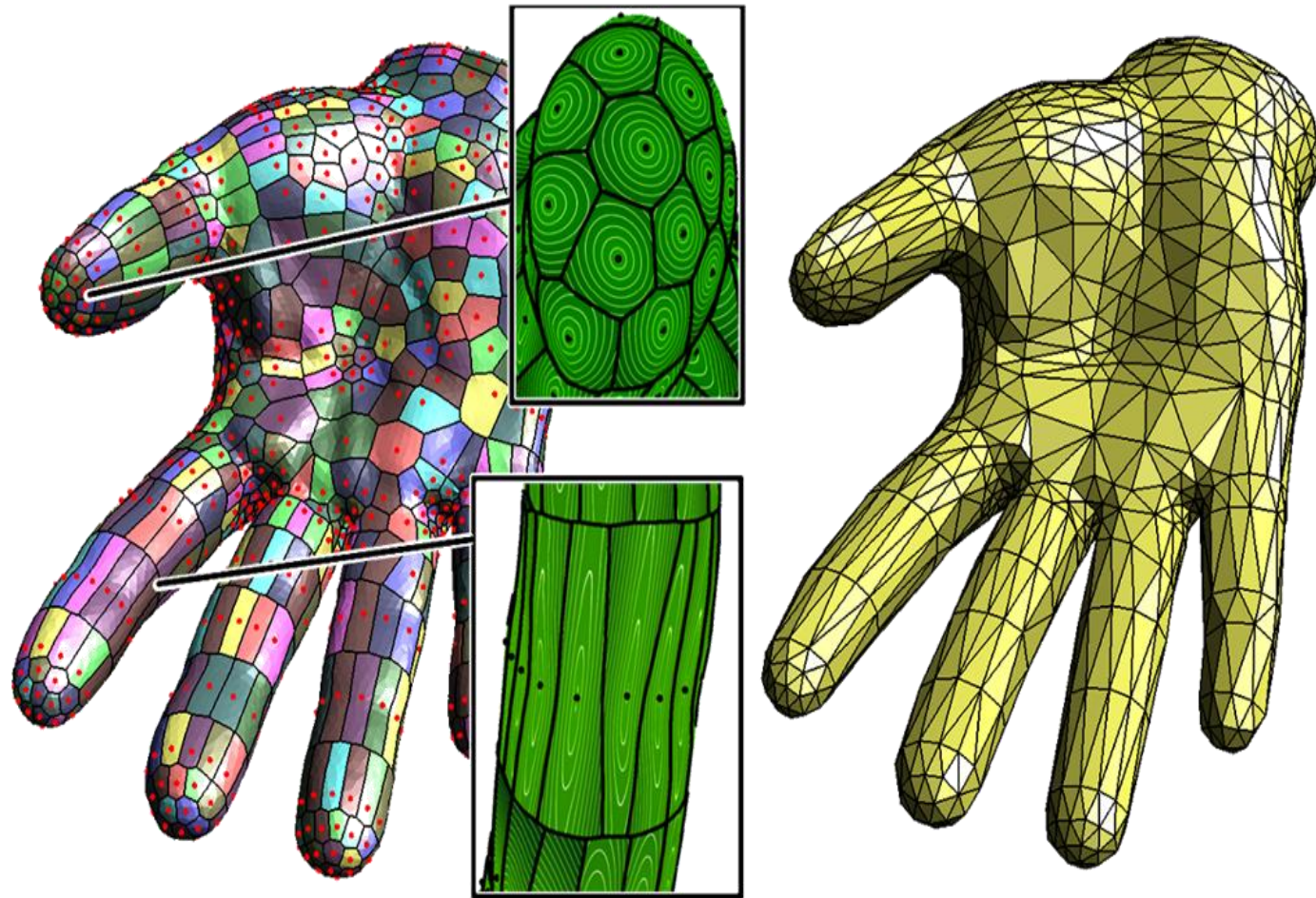
For each iterate  $\mathbf{X}^{(k)}$ :

Compute  $F(\mathbf{X}^{(k)})$  and  $\nabla F(\mathbf{X}^{(k)})$



# Zoom on Geometry Processing

## 4. Optimal Sampling

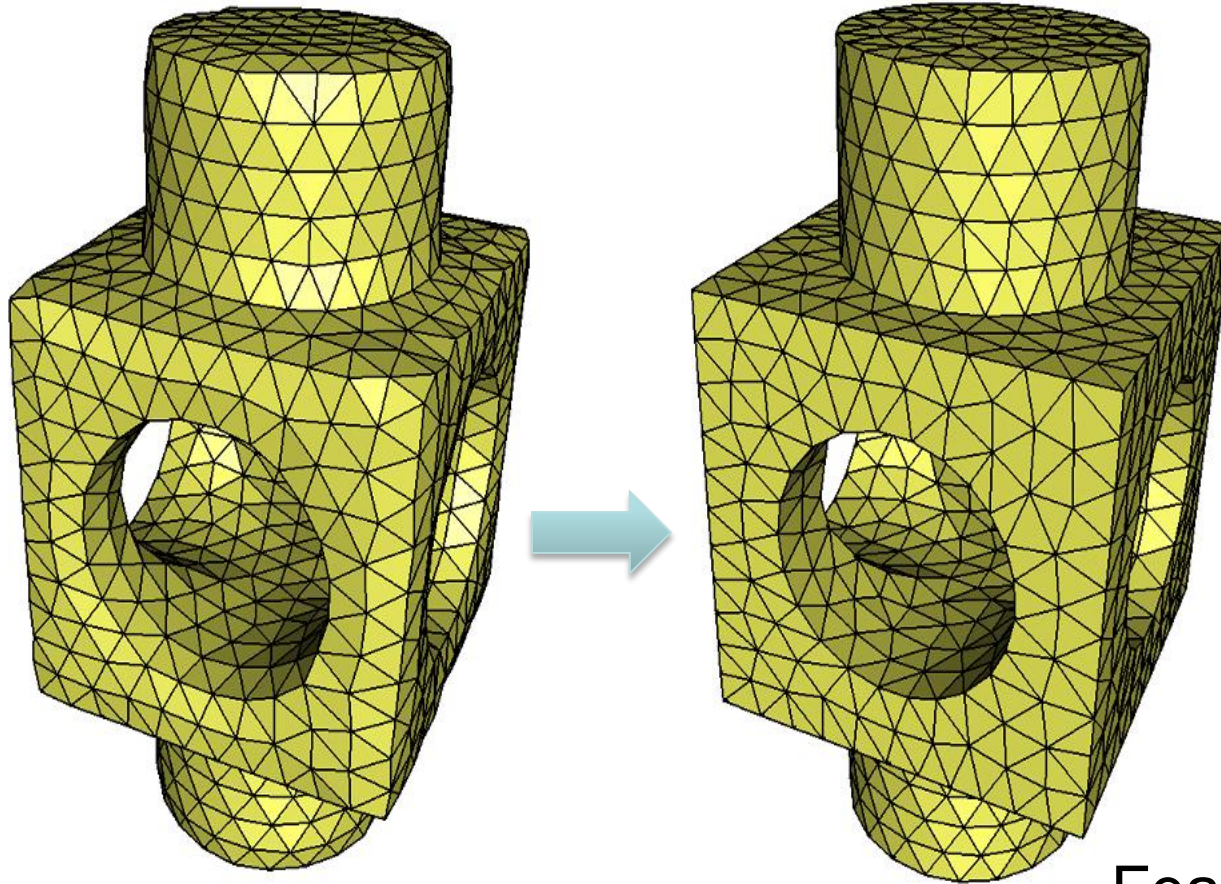


$p = 2$   
 $M(x) =$  ppal dir.  
of curvature.

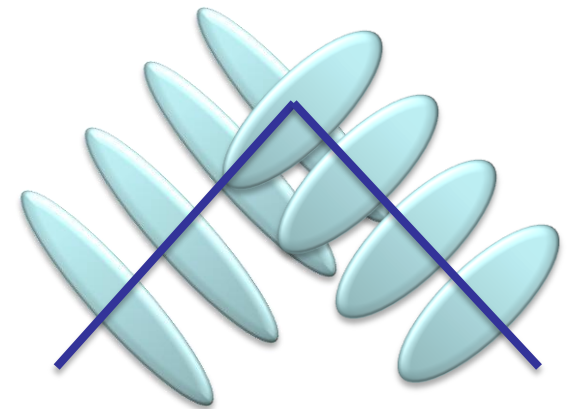


# Zoom on Geometry Processing

## 4. Optimal Sampling



$p = 2$   
 $M(x) = \text{Normal}$   
anisotropy.



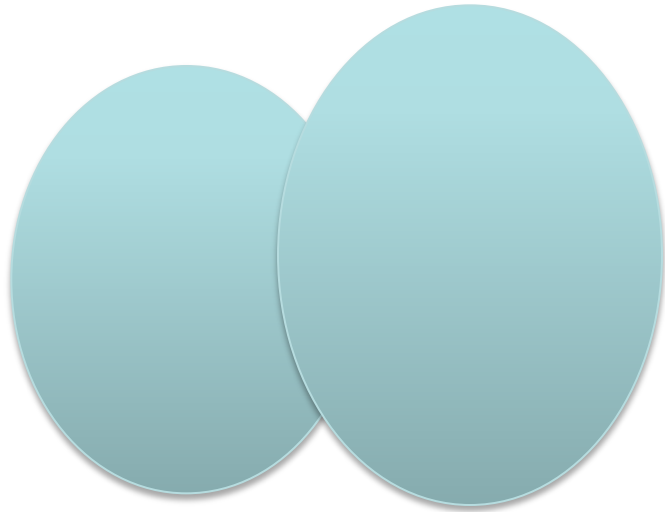
Feature-sensitive meshing





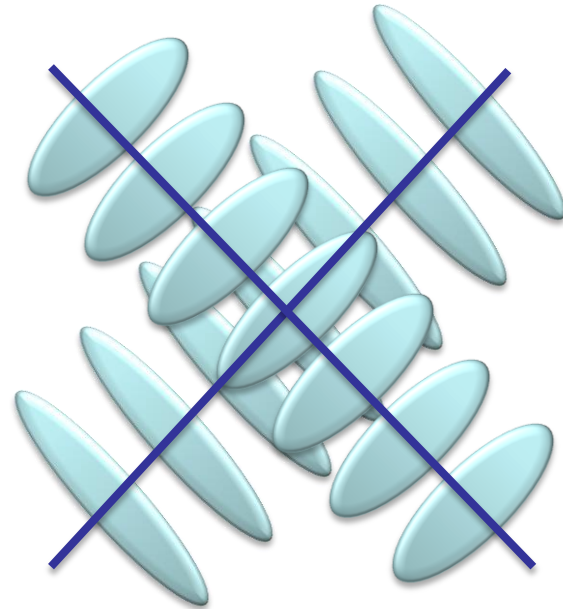
# Zoom on Geometry Processing

## 4. Optimal Sampling



CSG-Remeshing

$p = 2$   
 $M(x) = \text{Normal anisotropy.}$



Feature-sensitive meshing





# Zoom on Geometry Processing

## 4. Optimal Sampling



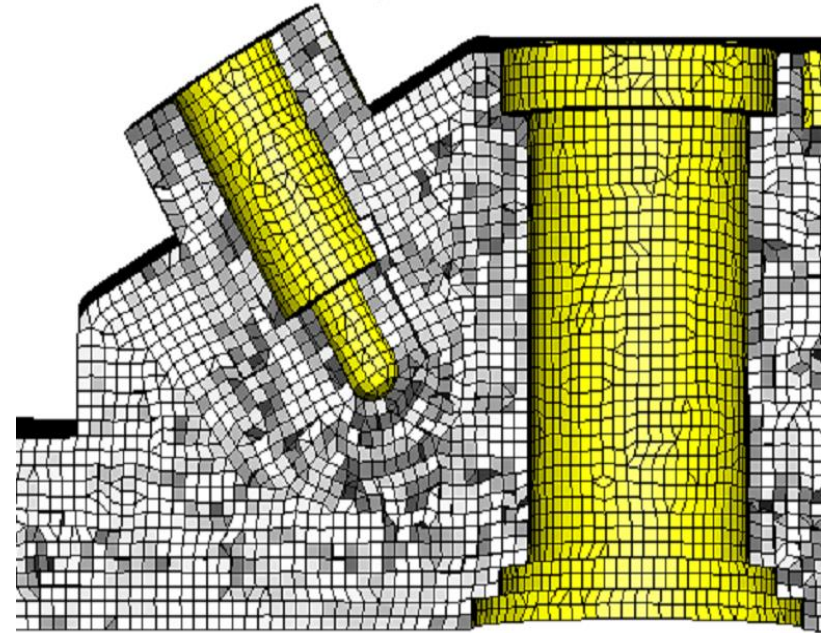
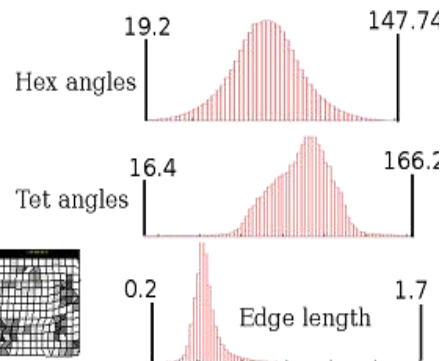
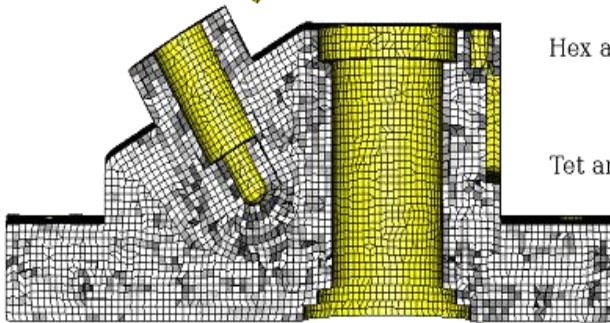
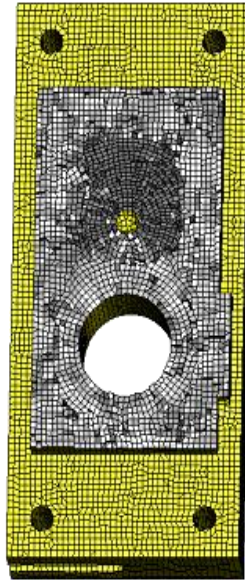
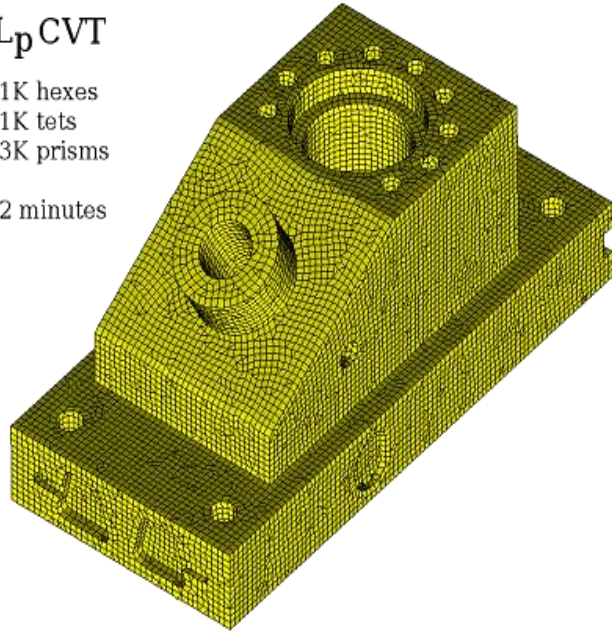
# Zoom on Geometry Processing

## 4. Optimal Sampling

$L_p$ CVT

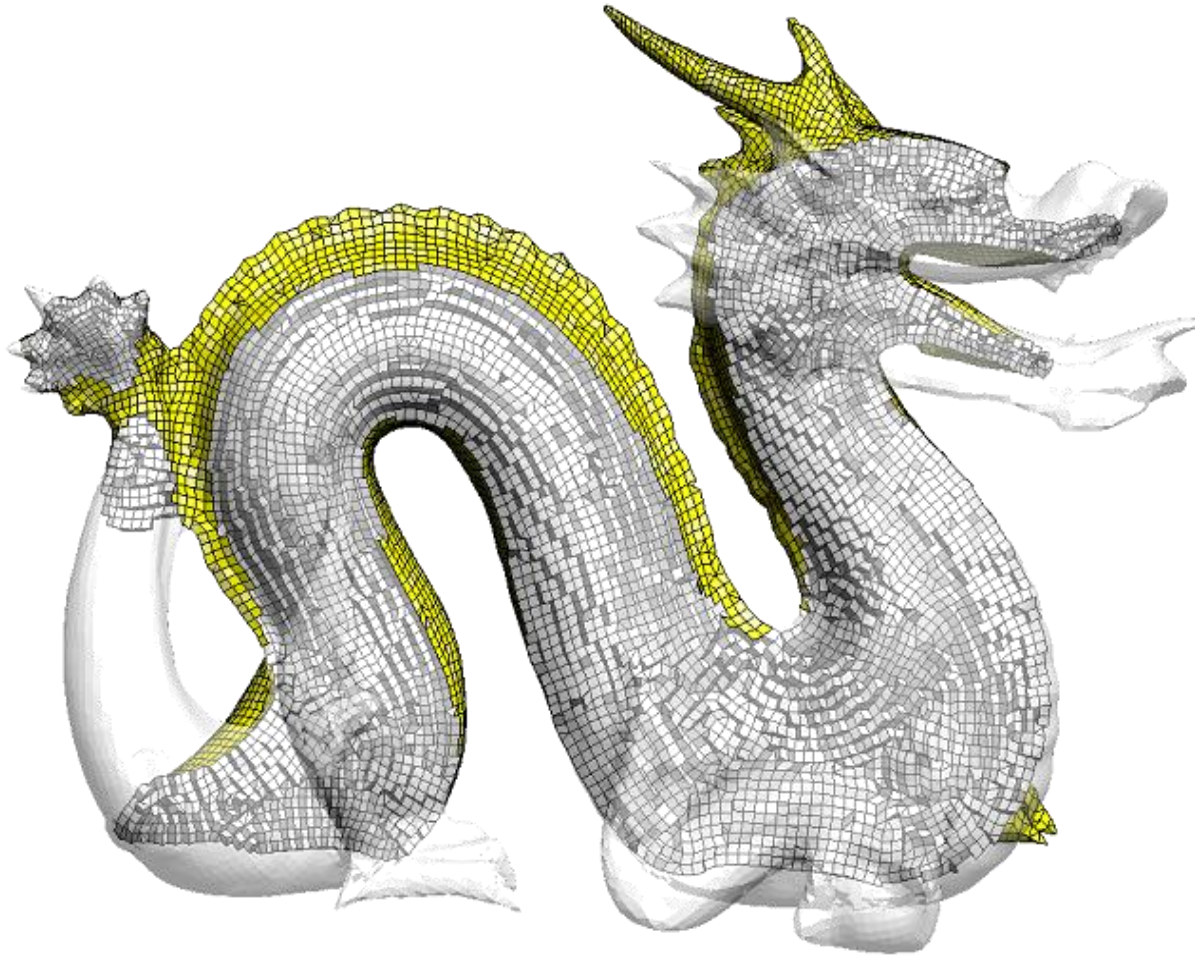
81K hexes  
11K tets  
13K prisms

12 minutes



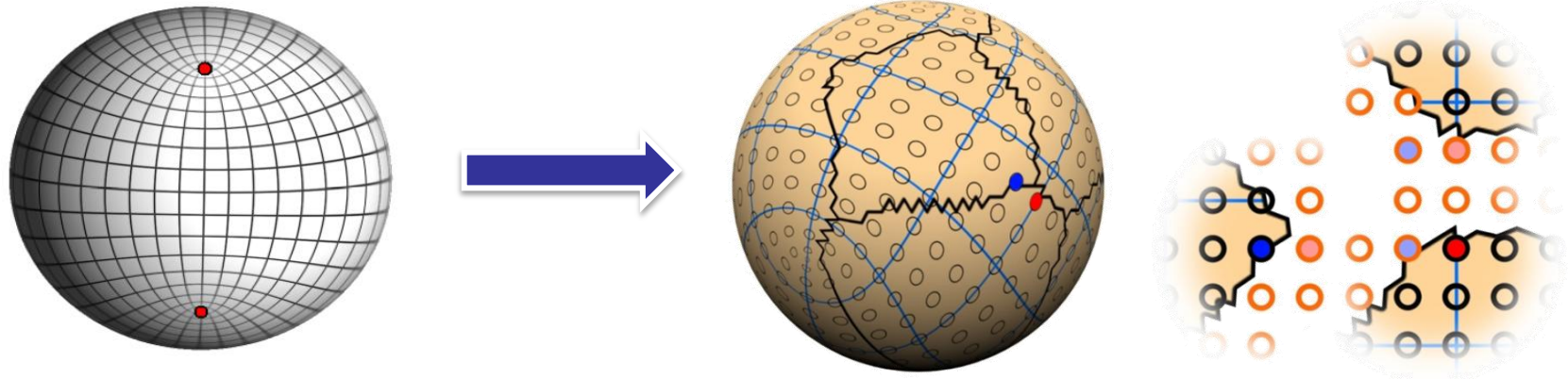
# Zoom on Geometry Processing

## 4. Optimal Sampling



# Zoom on Geometry Processing

## 5. Seamless texturing



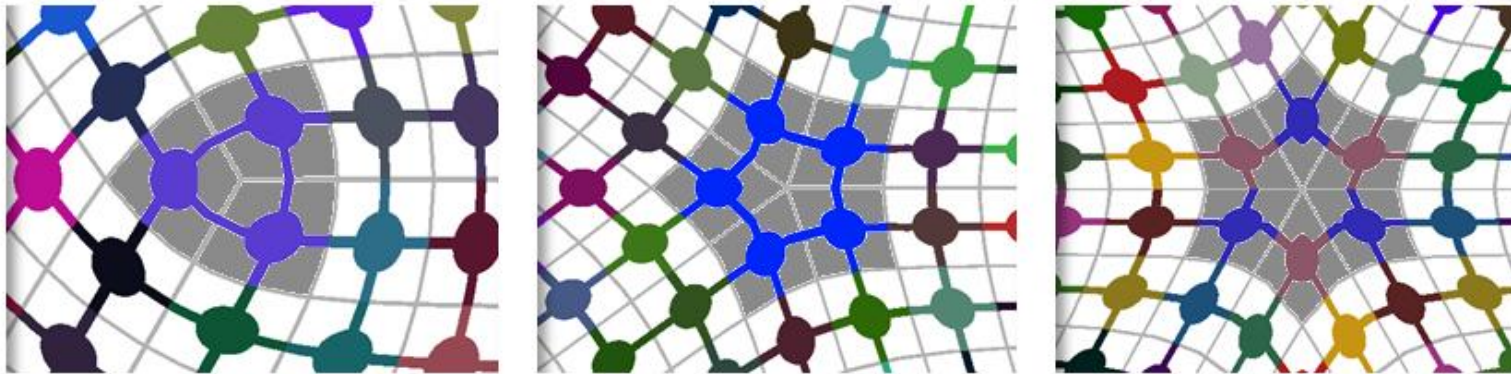
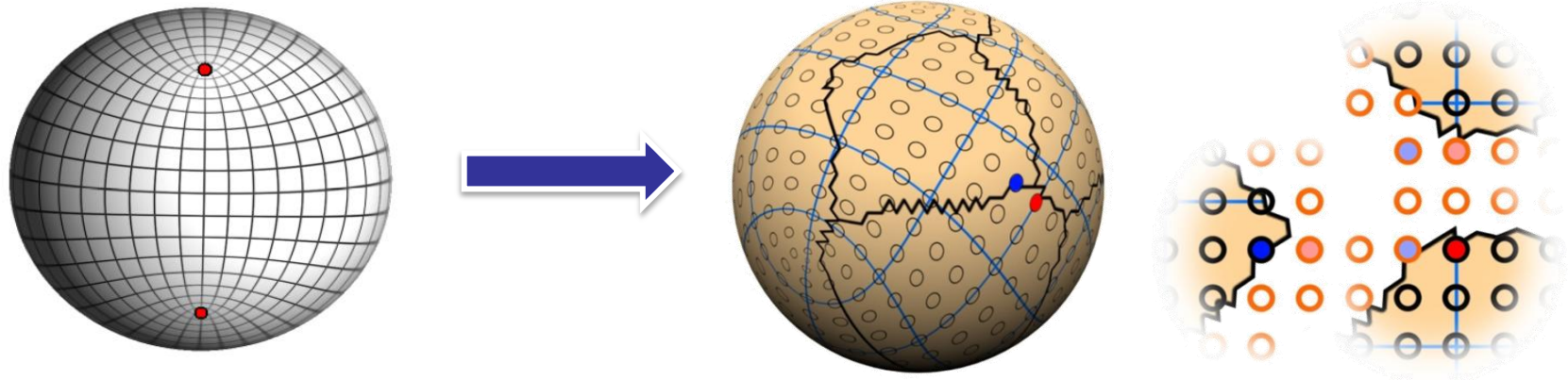
Invisible Seams [EGSR 2010]





# Zoom on Geometry Processing

## 5. Seamless texturing

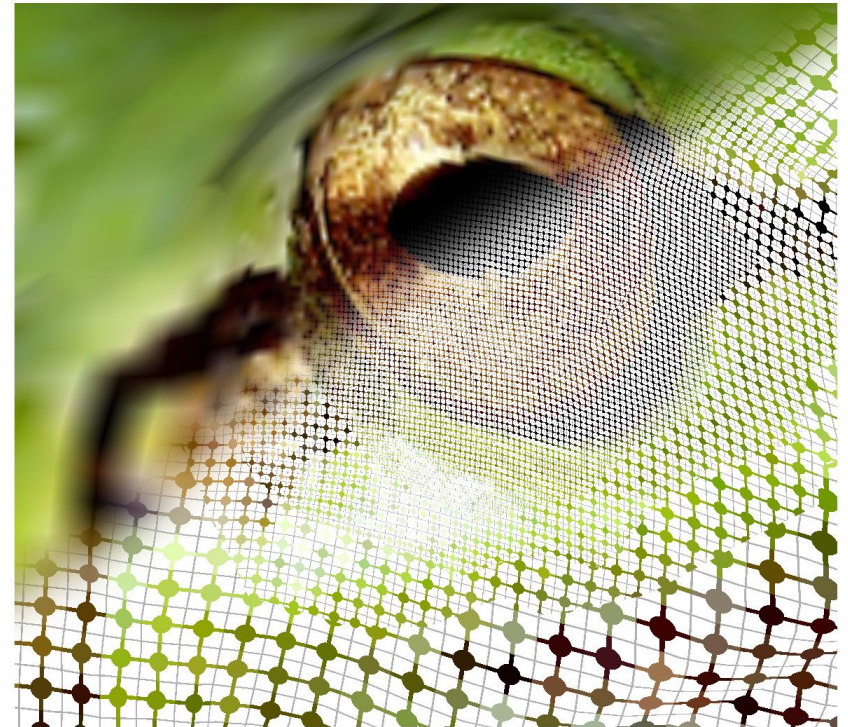


Invisible Seams [EGSR 2010]



# Zoom on Geometry Processing

## 5. Seamless Texturing



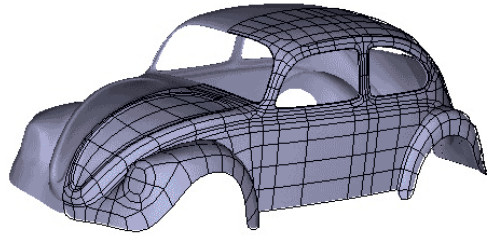
Invisible Seams [EGSR 2010]



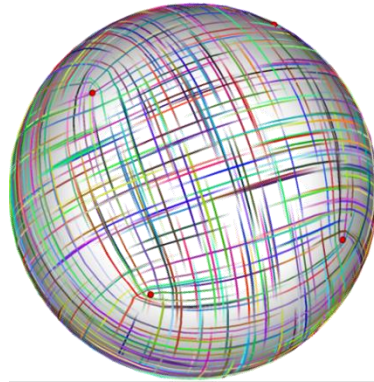


# Zoom on Geometry Processing

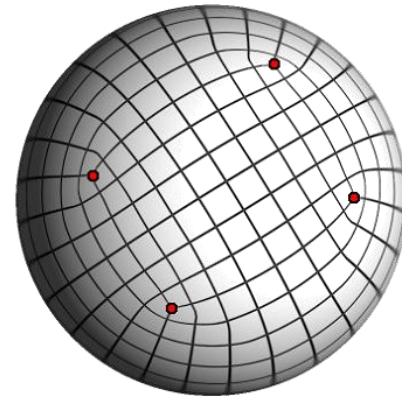
## Summary



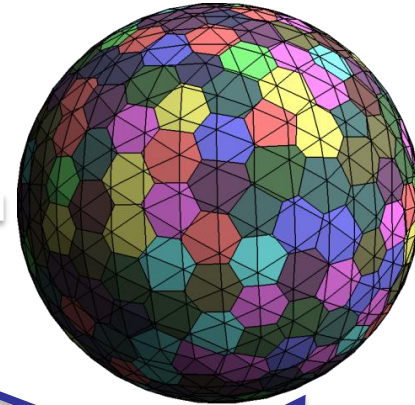
1. Intro  
Dynamic Function Basis



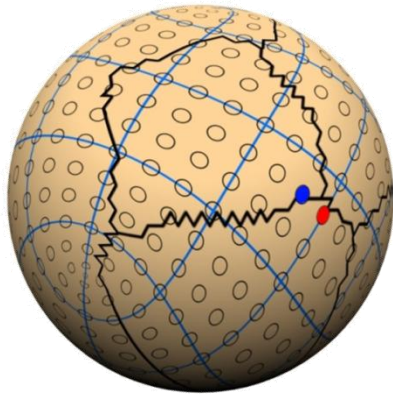
2. Direction Fields



3. Global Parameterization



4. Optimal Sampling ( $L_p$  and  $L_2$ )

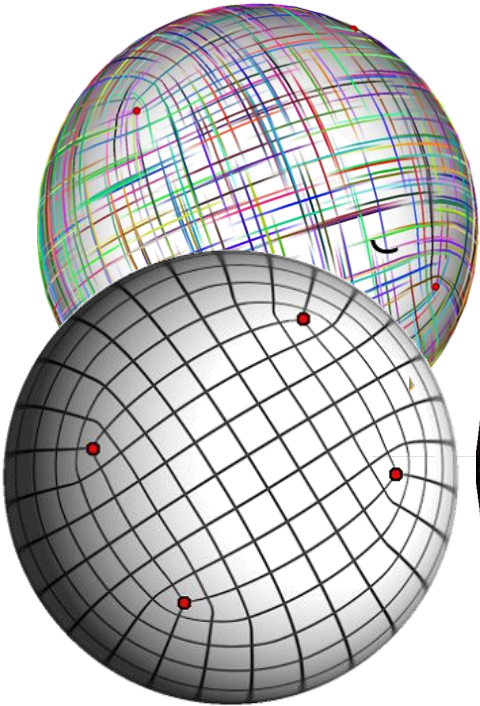


5. Seamless Texturing

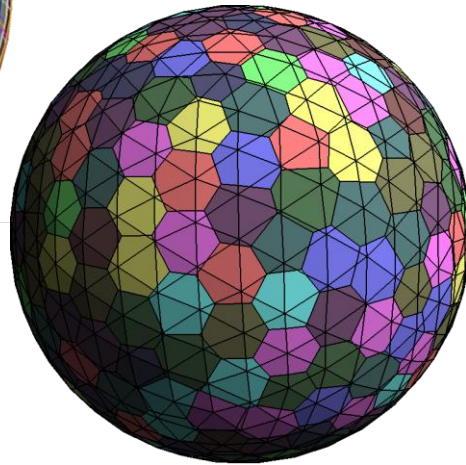


# Zoom on Geometry Processing

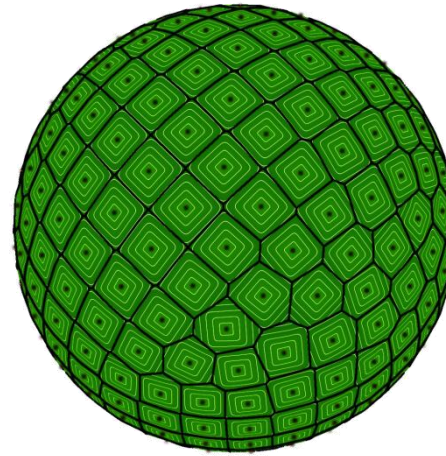
## Summary



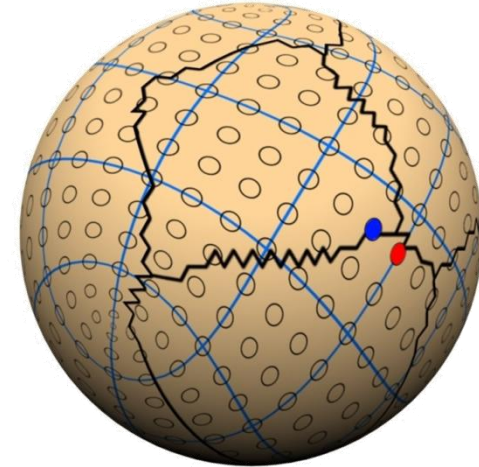
*Measuring*  
[ACM TOG 06,08]



*Sampling*  
[ACM TOG 09]



*Meshing*  
[SIGGRAPH 10]



*Mapping*  
[EGSR 10]





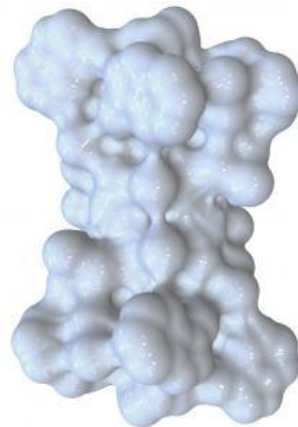
# Other works



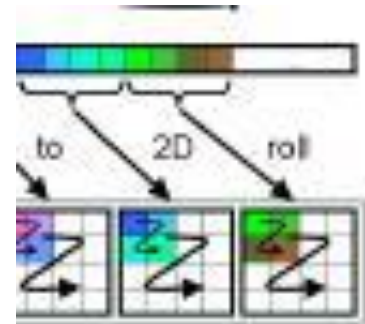
By-Example Synthesis of Architectural Textures, SIGGRAPH 2010  
Sylvain Lefebvre, Samuel Hornus and Anass Lasram  
(joint project with REVES)



Manifold Harmonics



Molecular Visualization

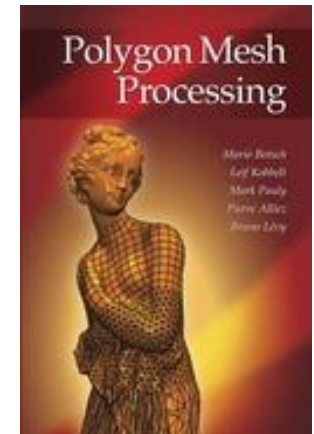


Concurrent Number Cruncher (GPU Solver)



# Impact Highlights

- \* 4 paper presentations + 1 course at SIGGRAPH 2010 (total talks time: 260 min.)
- \* Book « Polygon Mesh Processing » AK Peters
- \* Eurographics researcher's prize (S. Lefebvre)
- \* European Research Council grant and ANR Chaire d'Excellence
- \* Graphite most innovative special prize at « trophées du libre »



# Impact

## Visibility of Publications

(google scholar)

2008: Manifold Harmonics – 41 citations

2007: Concurrent Number Cruncher – 40 citations

2006: Periodic Global Parameterization – 90 citations



# Impact

## Visibility of Publications

(google scholar)

2008: Manifold Harmonics – 41 citations

2007: Concurrent Number Cruncher – 40 citations

2006: Periodic Global Parameterization – 90 citations

2002: Least Squares Conformal Maps – 504 citations





# Future Work – 2010-2014

## Dynamic Function Basis – Research Program

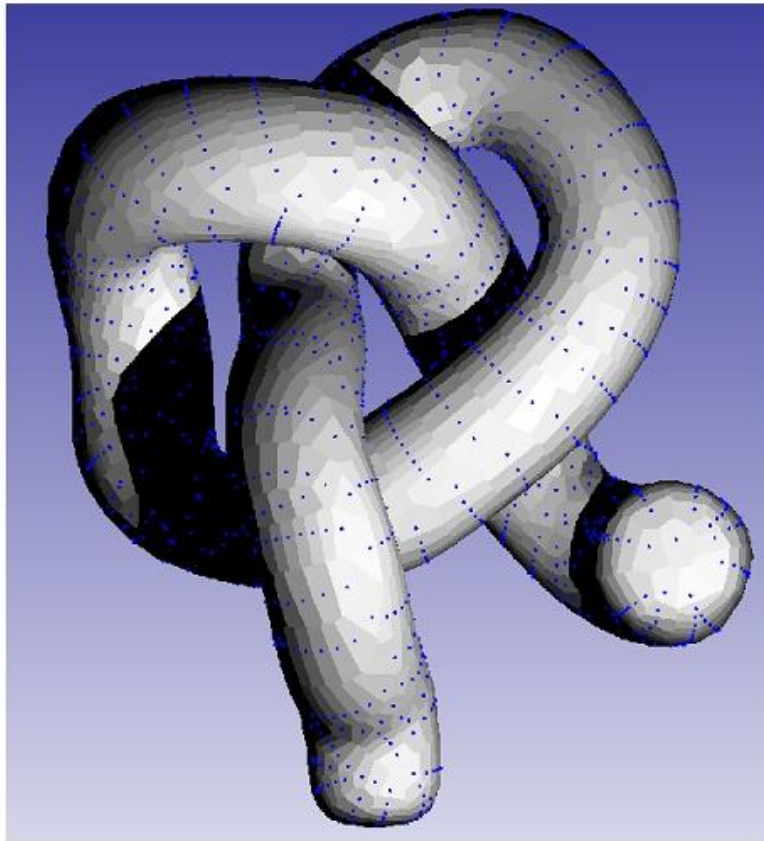
- 2D,  $L = Id$  : image approximation **[EGSR 2006]**
- 3D,  $L = Id$  : surface approximation **2006-2010**
- 3D, optimal sampling **2006-2010**
- 3D,  $L =$  light transport **2010-...**
- 3D+t, Navier Stokes, tracking **2010-...**

GOODSHAPE (ERC)  
PHYSIGRAPHICS (ANR)  
MORPHO (ANR)  
MODITERE (ANR)

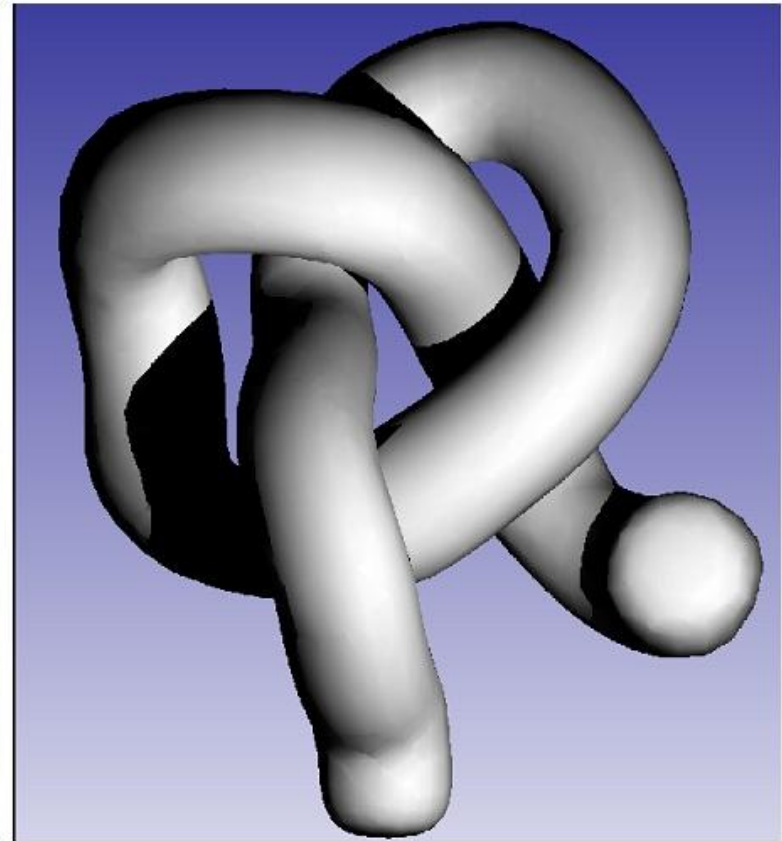


# Future Work – 2010-2014

## Dynamic Function Basis – Lighting



Dynamic Function Basis  
Constant elements

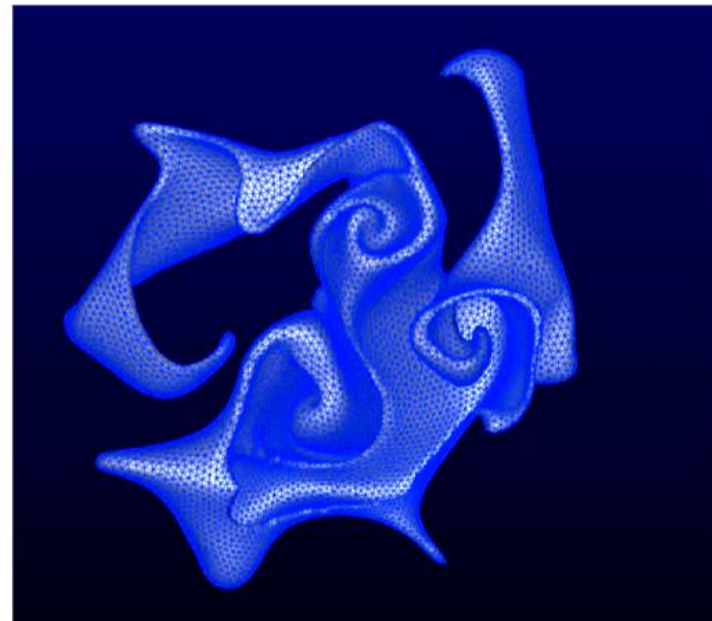
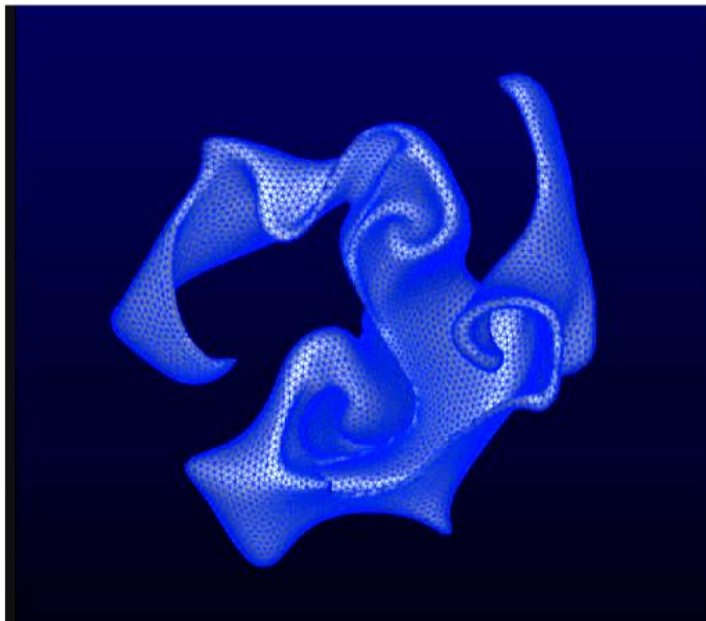
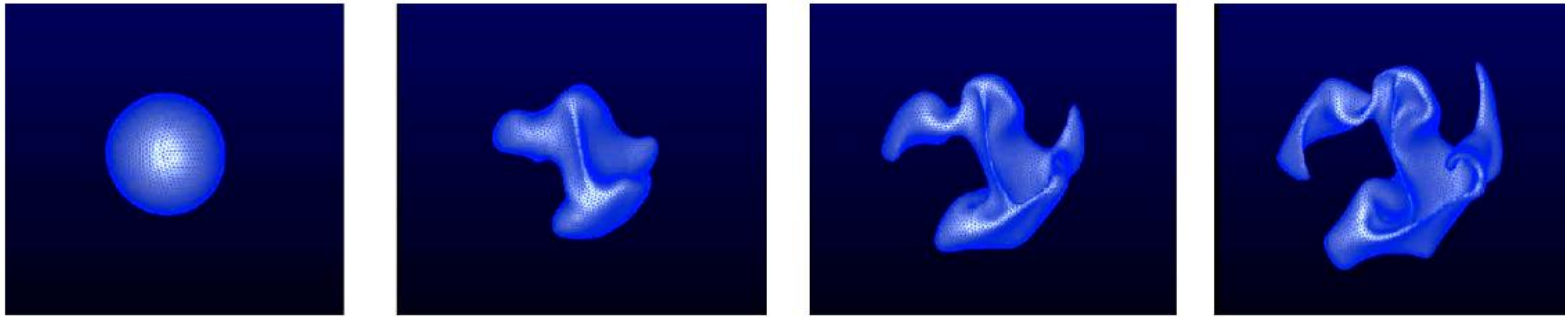


Dynamic Function Basis  
Quadratic elements



# Future Work – 2010-2014

## Dynamic Function Basis – Tracking



« Curlnoise » test



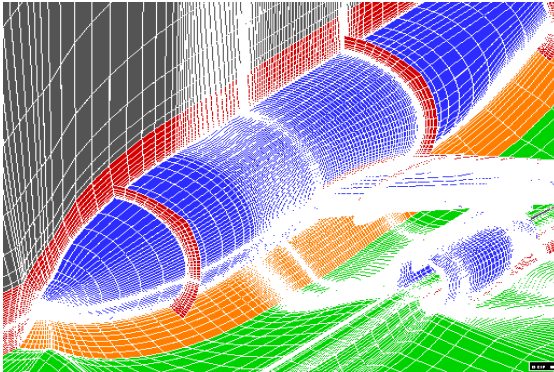
INSTITUT NATIONAL  
DE RECHERCHE  
EN INFORMATIQUE  
ET EN AUTOMATIQUE



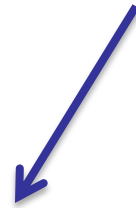
ALICE  
Geometry and Light

# Future Work – 2010-2014

Longer term ...(form bunnies to spaceships)



- 2D,  $L = Id$  : image approximation [EGSR 2006]
- 3D,  $L = Id$  : surface approximation 2006-2010
- 3D, optimal sampling 2006-2010
- 3D,  $L =$  light transport 2010-...
- 3D+t, Navier Stokes, tracking 2010-...
- **Finite Elements Modeling** 2012-...



Connections with Applied Mathematics community

Wider project: New Foundations for Numerical Engineering





# Future Work – 2010-2014

## New ALICE research directions

### ▼ Applied Mathematics

Finite Element Modeling  
Numerics  
Differential Geometry  
Computational Physics



### ■ Content Creation

By-Example Modeling  
Texturing  
Geometry Synthesis



# Future Work – 2010-2014

## New ALICE research directions

### ▼ Applied Mathematics

Finite Element Modeling  
Numerics  
Differential Geometry  
Computational Physics

### ■ Content Creation

By-Example Modeling  
Texturing  
Geometry Synthesis



Thank you for your attention

