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Some advances in open problems of isogeometric analysis

M. Brovka^{(1)*}, J.I. López⁽¹⁾, J.M. Escobar⁽¹⁾, J.M. Cascón⁽²⁾ and R. Montenegro⁽¹⁾

⁽¹⁾ University Institute SIANI, University of Las Palmas de Gran Canaria, Spain

⁽²⁾ Department of Economics and History of Economics, University of Salamanca, Spain

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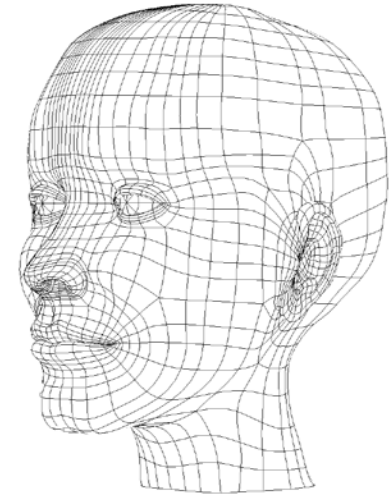
MINECO y FEDER Project: CGL2011-29396-C03-00

CONACYT-SENER Project, Fondo Sectorial, contract: 163723

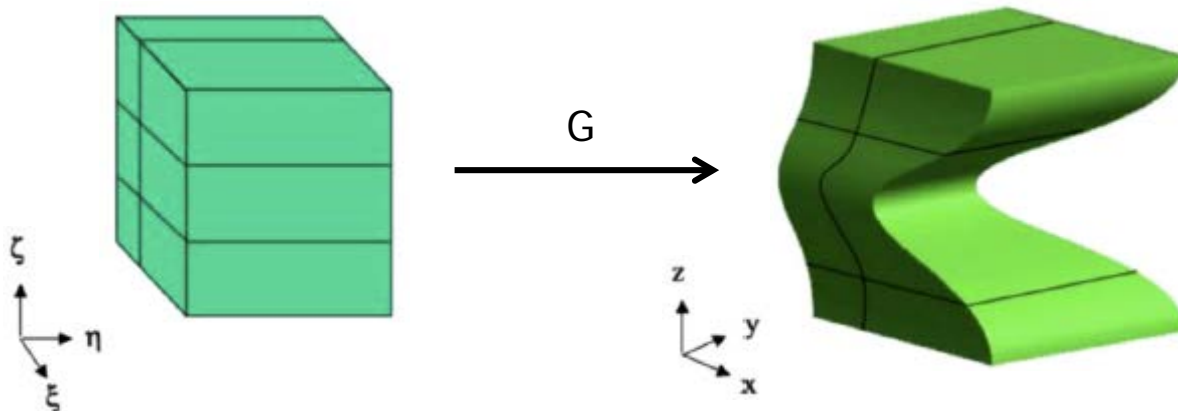
<http://www.dca.iusiani.ulpgc.es/proyecto2012-2014>

Isogeometric analysis (IGA) has arisen as an attempt to unify the field of CAD and classical finite element method

The main idea: using for analysis the same functions that are used in CAD representation of the computational domain, preserving thus the “exact” geometry



Global parameterization of the physical domain



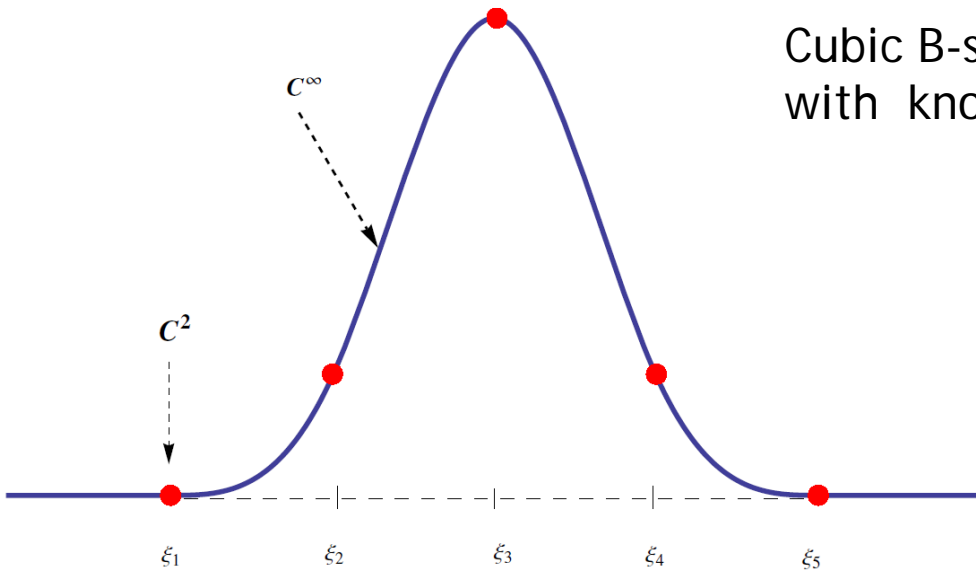
Basis functions: cubic splines

Higher smoothness:
 C^2 instead of C^0 of FEM

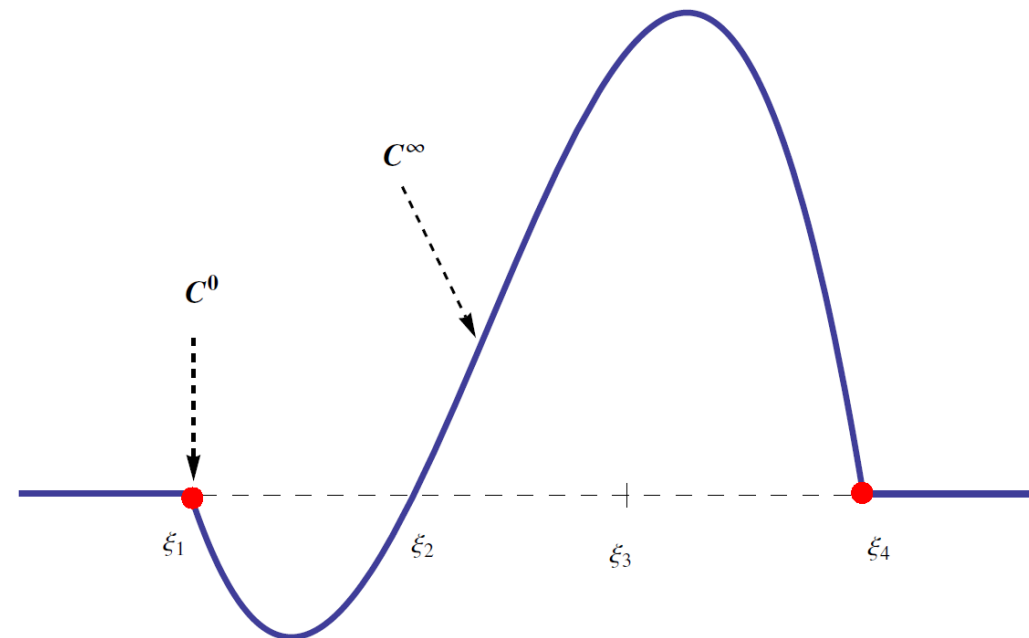
Isogeometric analysis: integrating FEA and CAD

B-spline basis functions

Cubic B-spline
with knot vector $\{\xi_1, \xi_2, \xi_3, \xi_4, \xi_5\}$



Langrange basis function of order 3



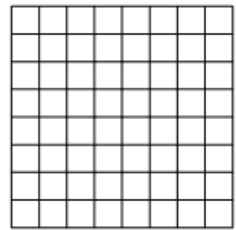
Isogeometric analysis: integrating FEA and CAD

Isoparametric approach in IGA

(T. Hughes et al. 2005) uses **only one global geometry function**

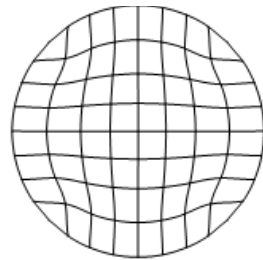
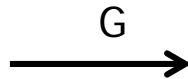
$$G : \Omega_0 = [0, 1]^2 \rightarrow \Omega$$

$$G(\xi) = \sum_i R_i(\xi) \mathbf{d}_i \text{ with tensor-product NURBS basis functions } R_i$$



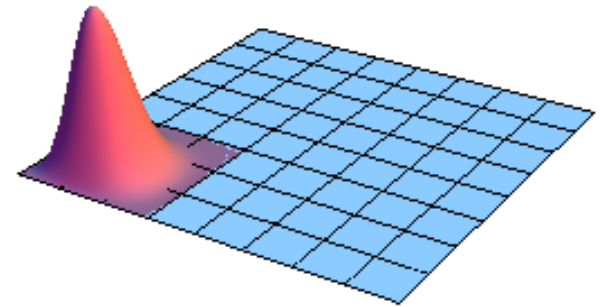
$$\xi \in \Omega_0$$

$$(\xi = (\xi, \eta))$$

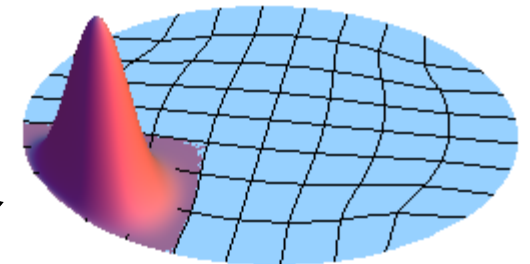


$$x(\xi) \in \Omega$$

$$R_i$$



$$\phi_i$$

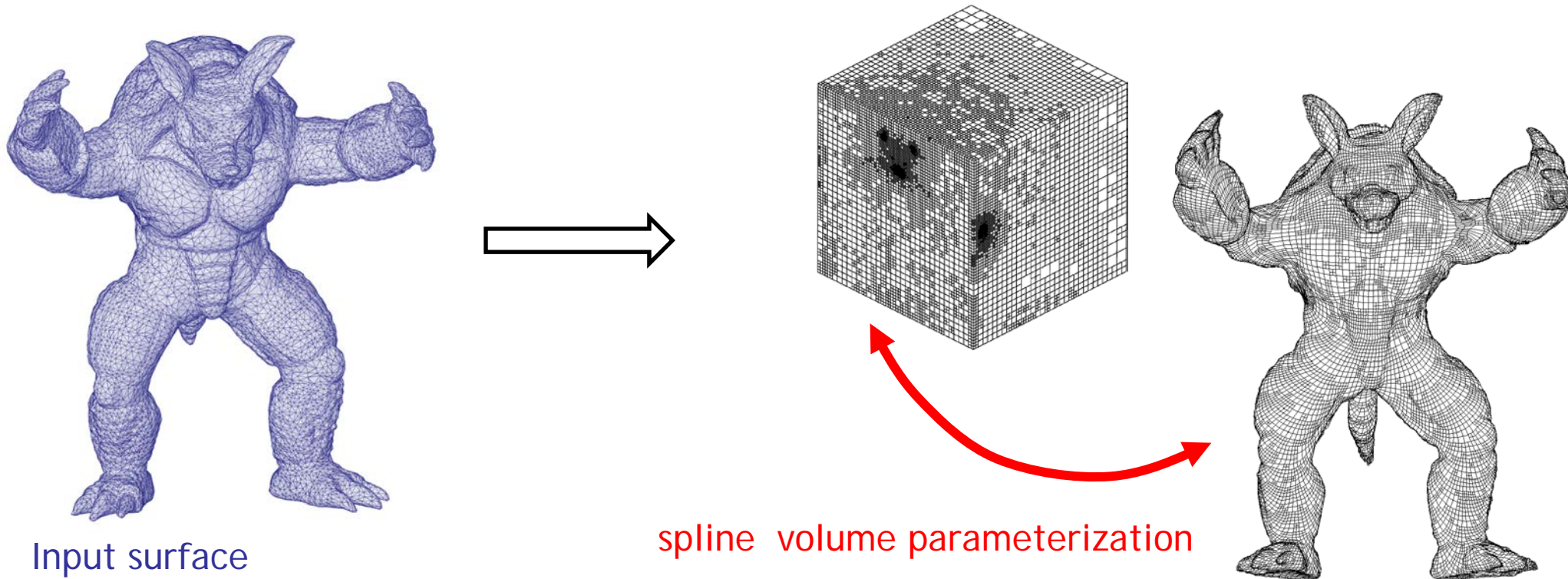


The basis functions of $V_h \subset V$ are $\phi_i = R_i \circ G^{-1}$

Open problems

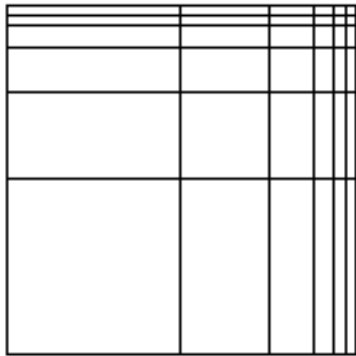
- Volumetric parameterization of computational domain from its surface representation.

CAD provides only surface representation of the geometry. For application of IGA it is necessary to have robust and effective method to obtain analysis-suitable volume parameterization

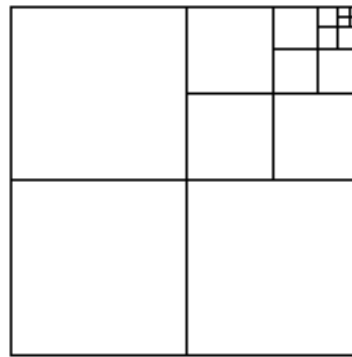


□ Local refinement.

Tensor product structure does not allow local enrichment of approximation space: knot insertion propagates thru the domain. A strategy for defining spline spaces over meshes with T-junctions (T-meshes) is needed.



global refinement



local refinement

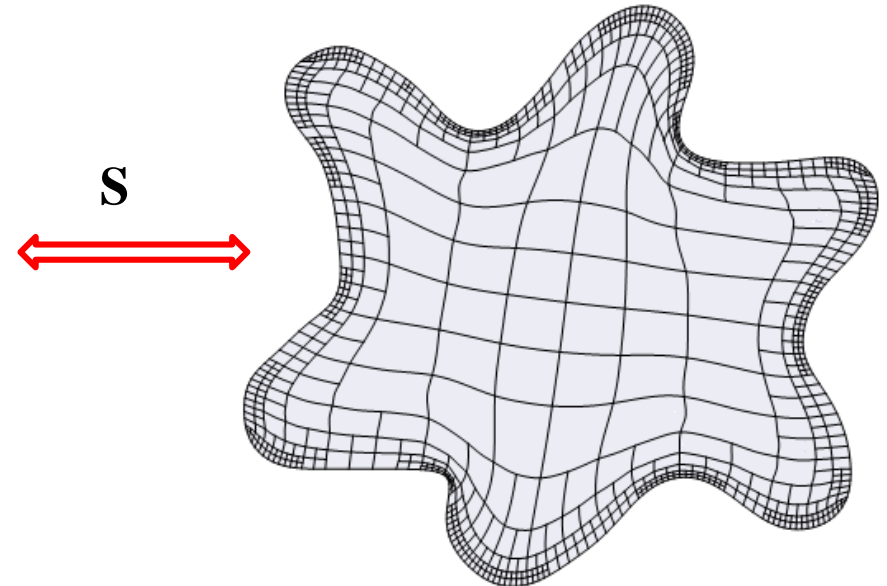
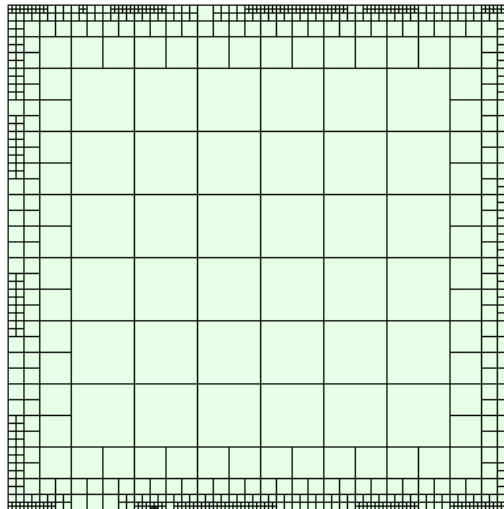
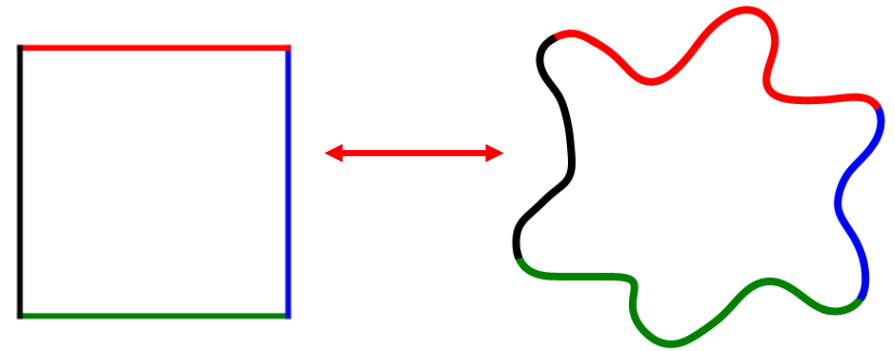
- A method for spline parameterization of 2D and 3D geometries
- A new strategy for constructing cubic spline spaces over quadtree (2D) and octree (3D) T-meshes

Our results: spline parameterization of 2D geometries

Goal: construct a global transformation from parametric to physical domain from boundary representation of the geometry

Good quality parameterization:

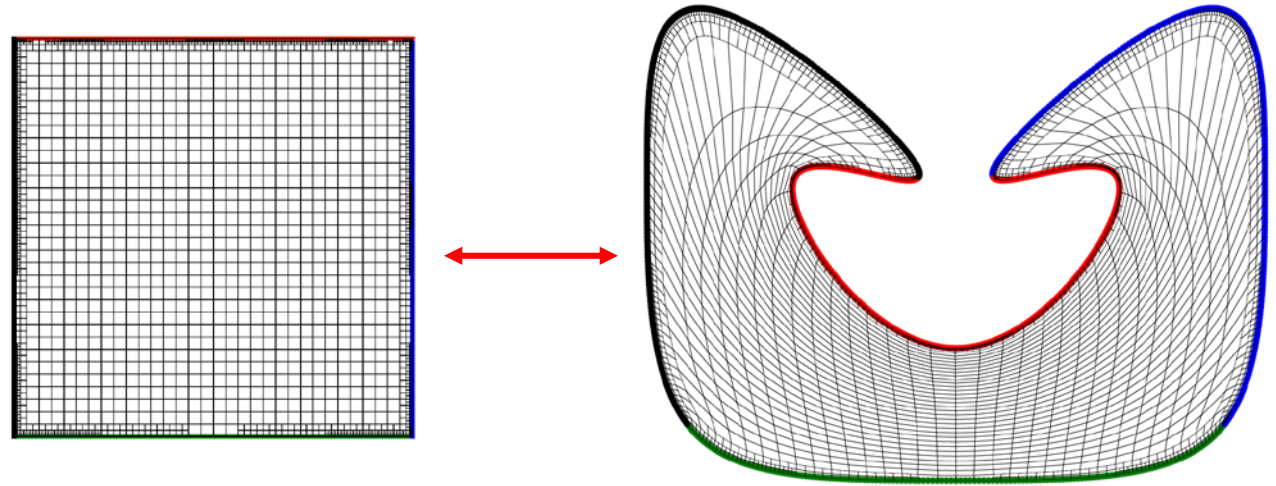
- Strictly positive Jacobian
- Good orthogonality and uniformity of isoparametric curves



Spline parameterization of 2D geometries

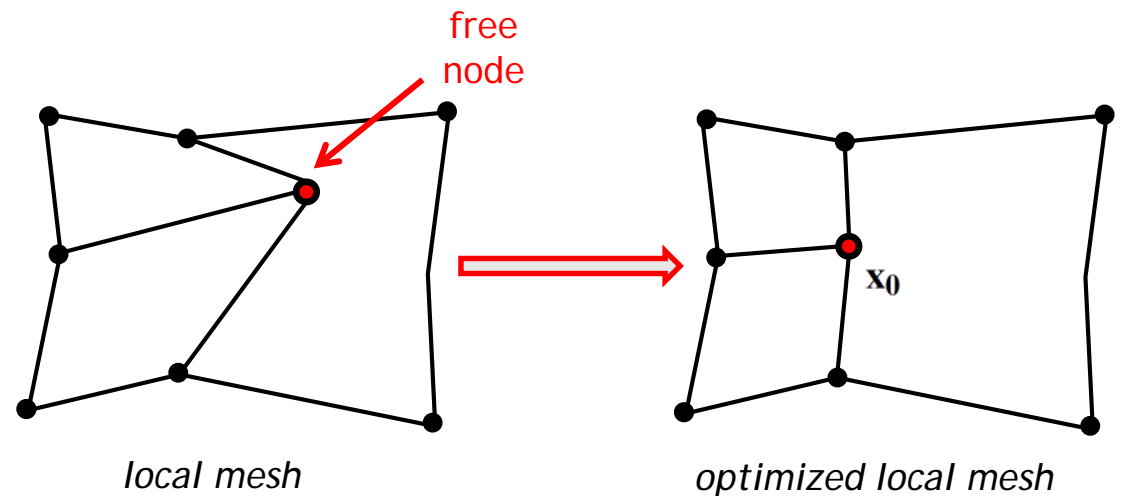
The key: untangling and smoothing procedure for T-mesh

Parametric T-mesh is deformed isomorphically into the physical T-mesh



Local optimization: determine a new position of the free node to improve mesh quality

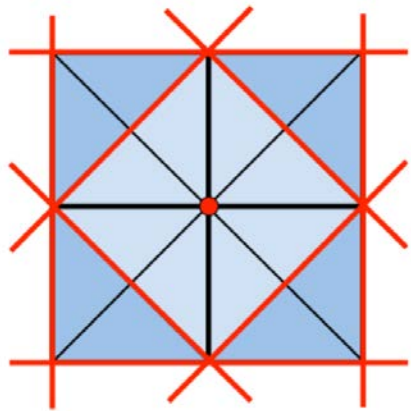
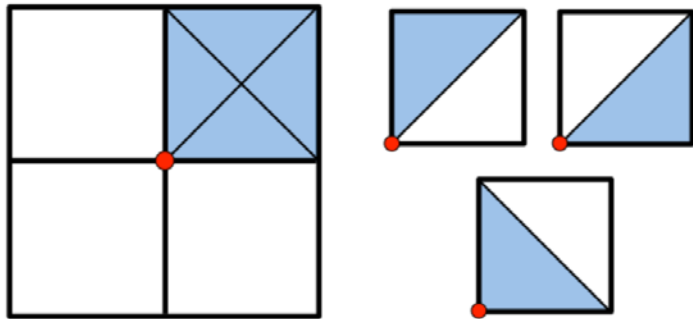
Minimize the objective function $K(\mathbf{x})$ to find the optimal position \mathbf{x}_0 of the free node



Simultaneous Untangling and Smoothing of T-meshes

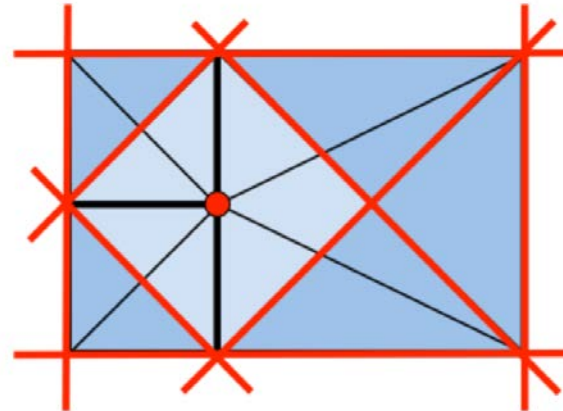
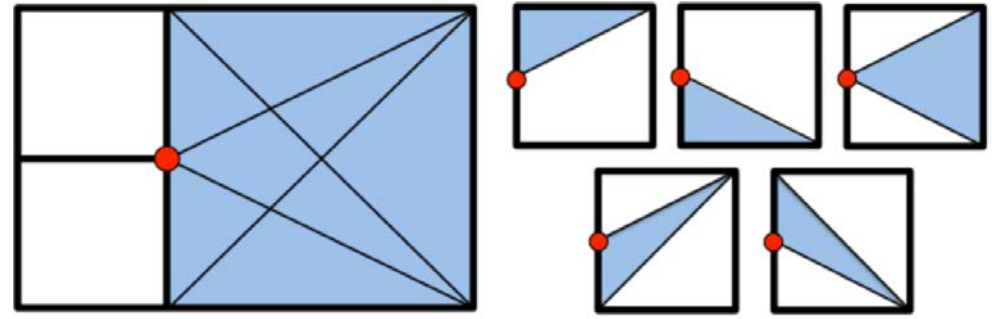
Triangle decomposition of the T-mesh cells

Case 1: Free node is a regular node



Barriers and feasible region for a regular node

Case 2: Free node is a hanging node



Barriers and feasible region for a hanging node

Simultaneous Untangling and Smoothing of T-meshes

Quality measure. Objective function

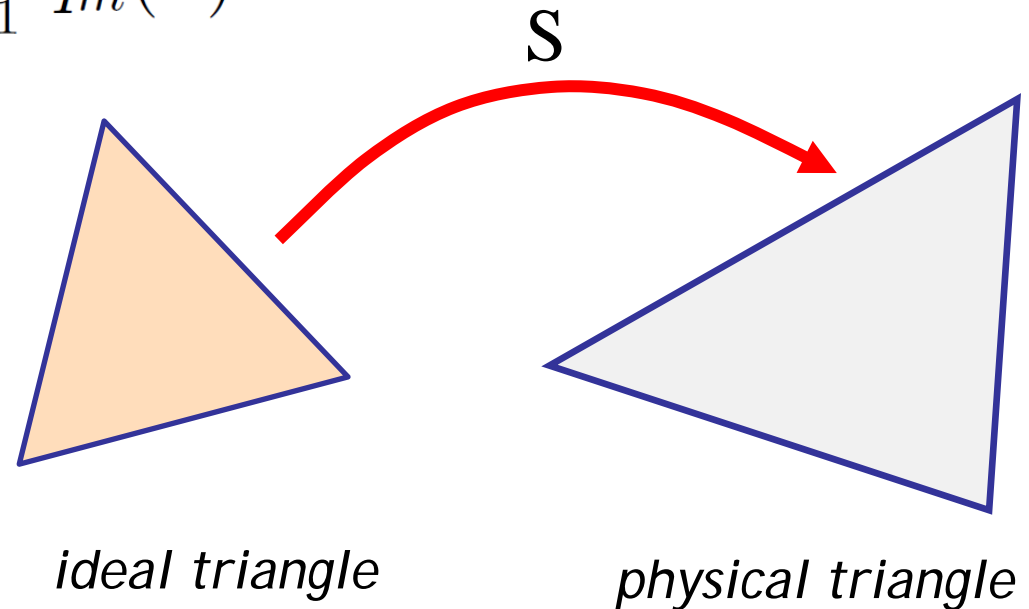
Local submesh is decomposed in triangles.

Shape quality measure for a triangle: *mean ratio* $q = \frac{2 \det(S)}{\|S\|^2}$

Objective function: $K(\mathbf{x}) = \sum_{m=1}^M \frac{1}{q_m(\mathbf{x})}$

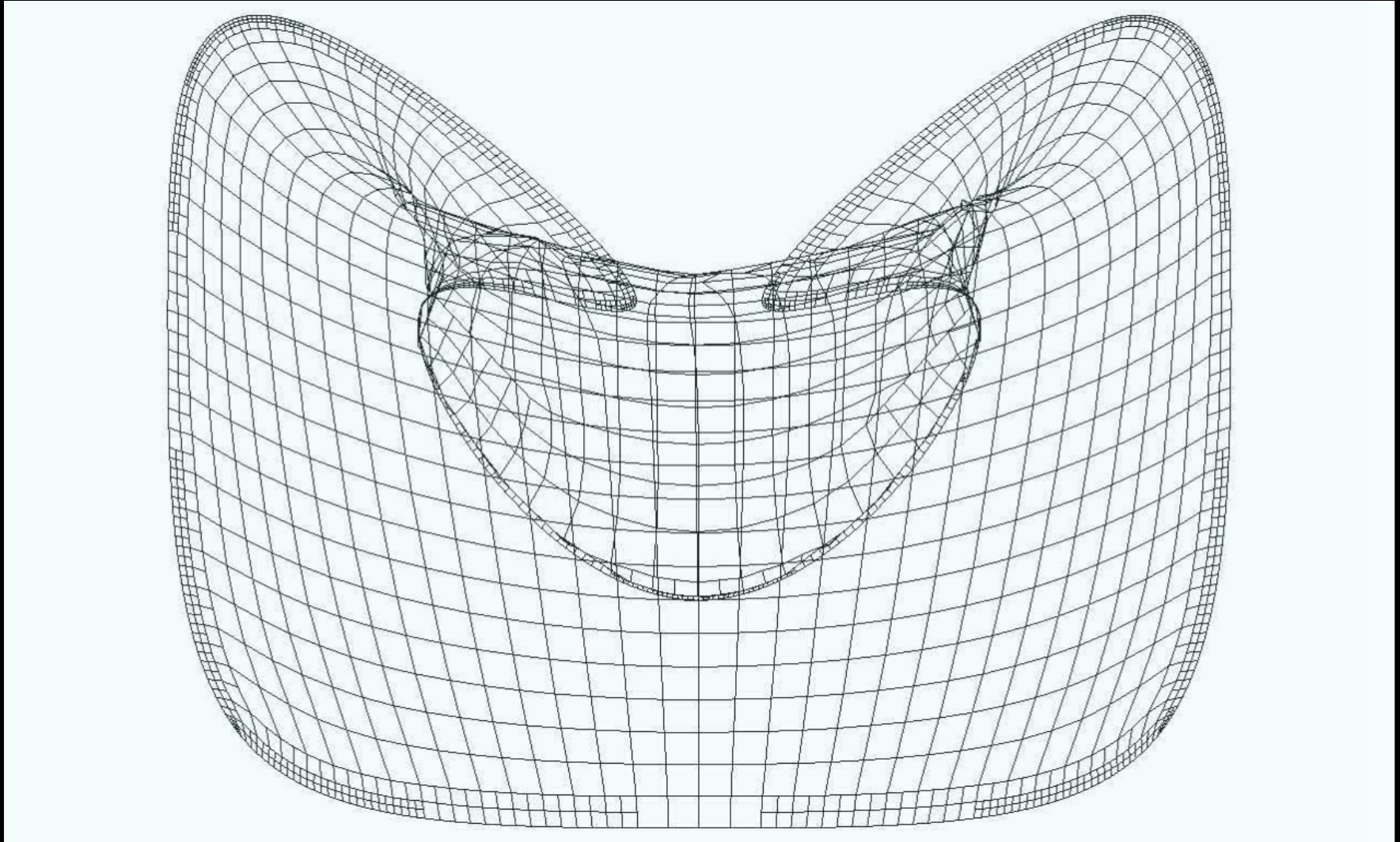
$$K(\mathbf{x}) = \sum_{m=1}^M \frac{\|S_m\|^2}{2 \det(S_m)}$$

M: number of elements of the local submesh



Simultaneous Untangling and Smoothing of T-meshes

T-mesh transformation along the SUS process: Video

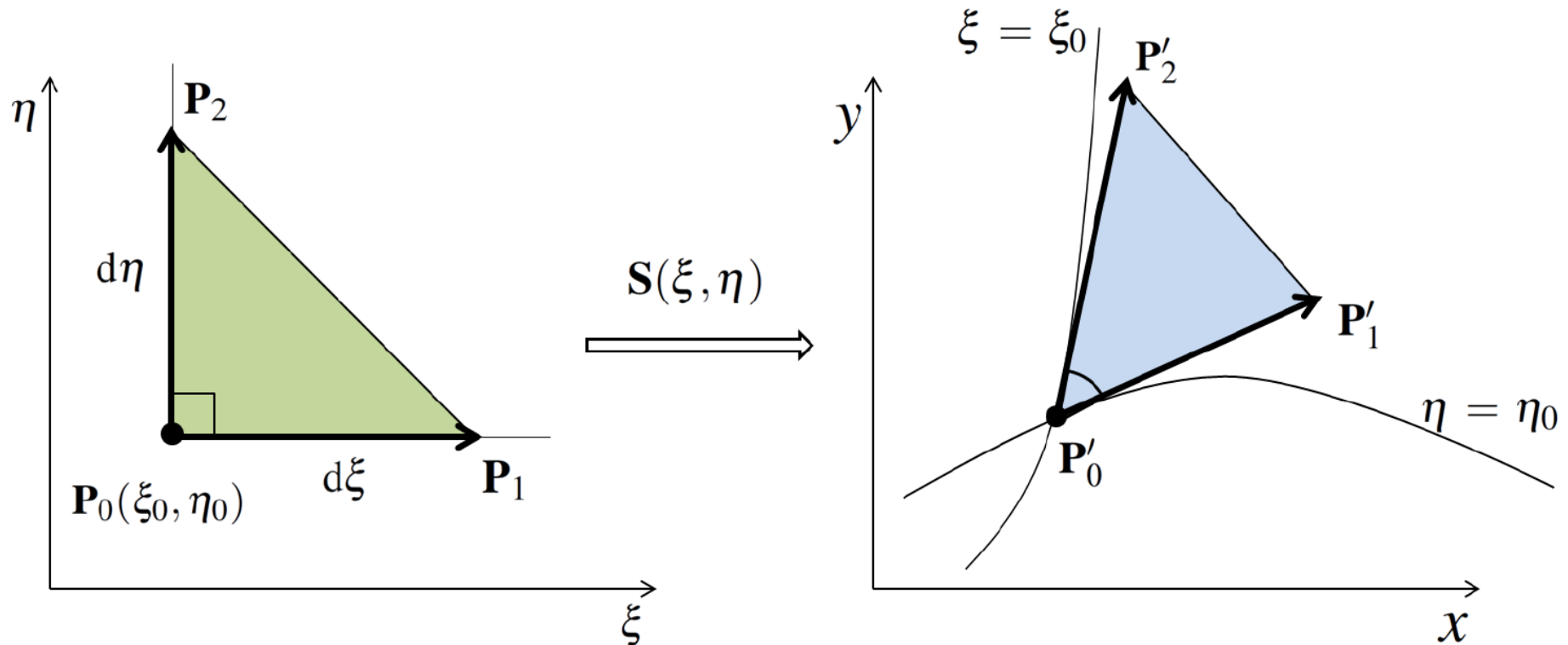


Mean Ratio Jacobian $J_r(\xi)$ of Parametric Transformation

A quality metric of the mapping at any point P_0

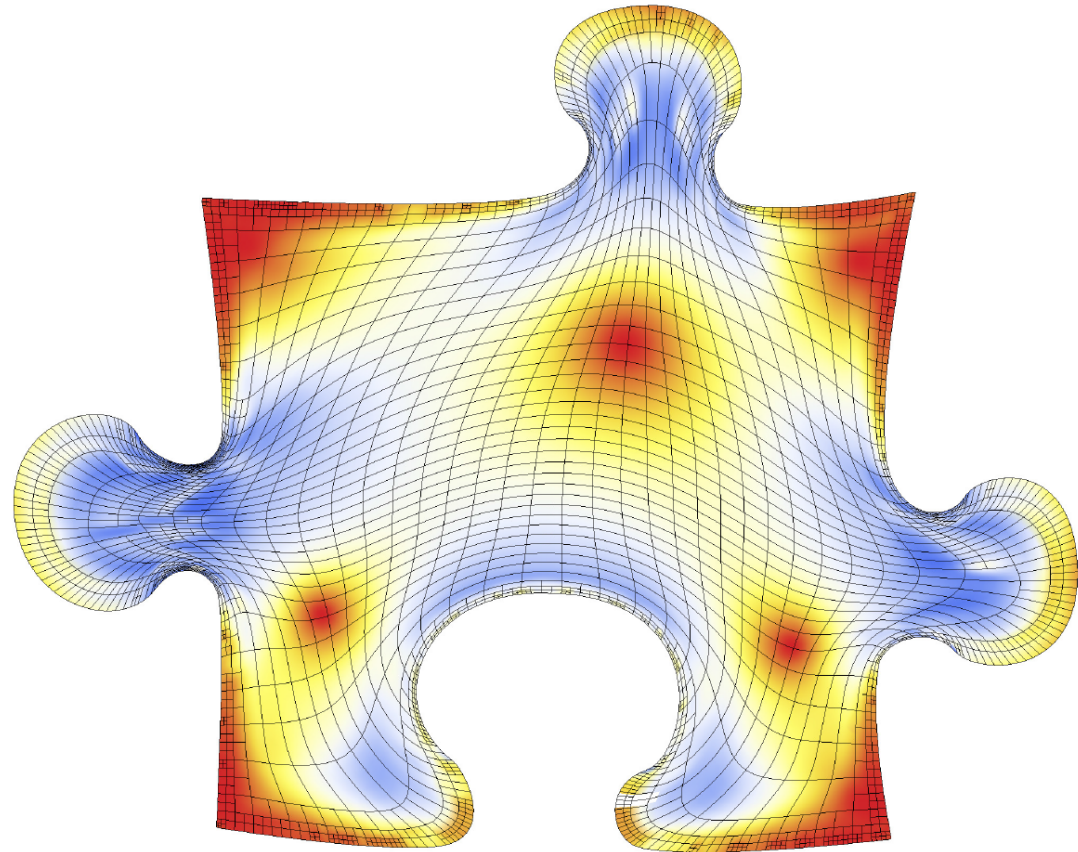
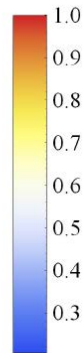
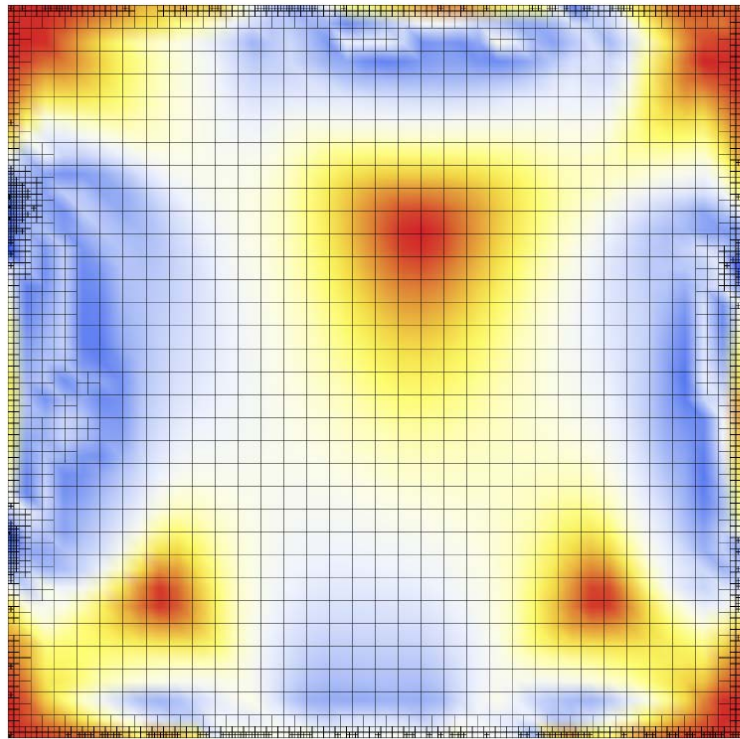
$$-1 \leq J_r(\xi) = \frac{2 \det(J)}{\|J\|^2} \leq 1$$

where J is the jacobian matrix of the spline mapping S



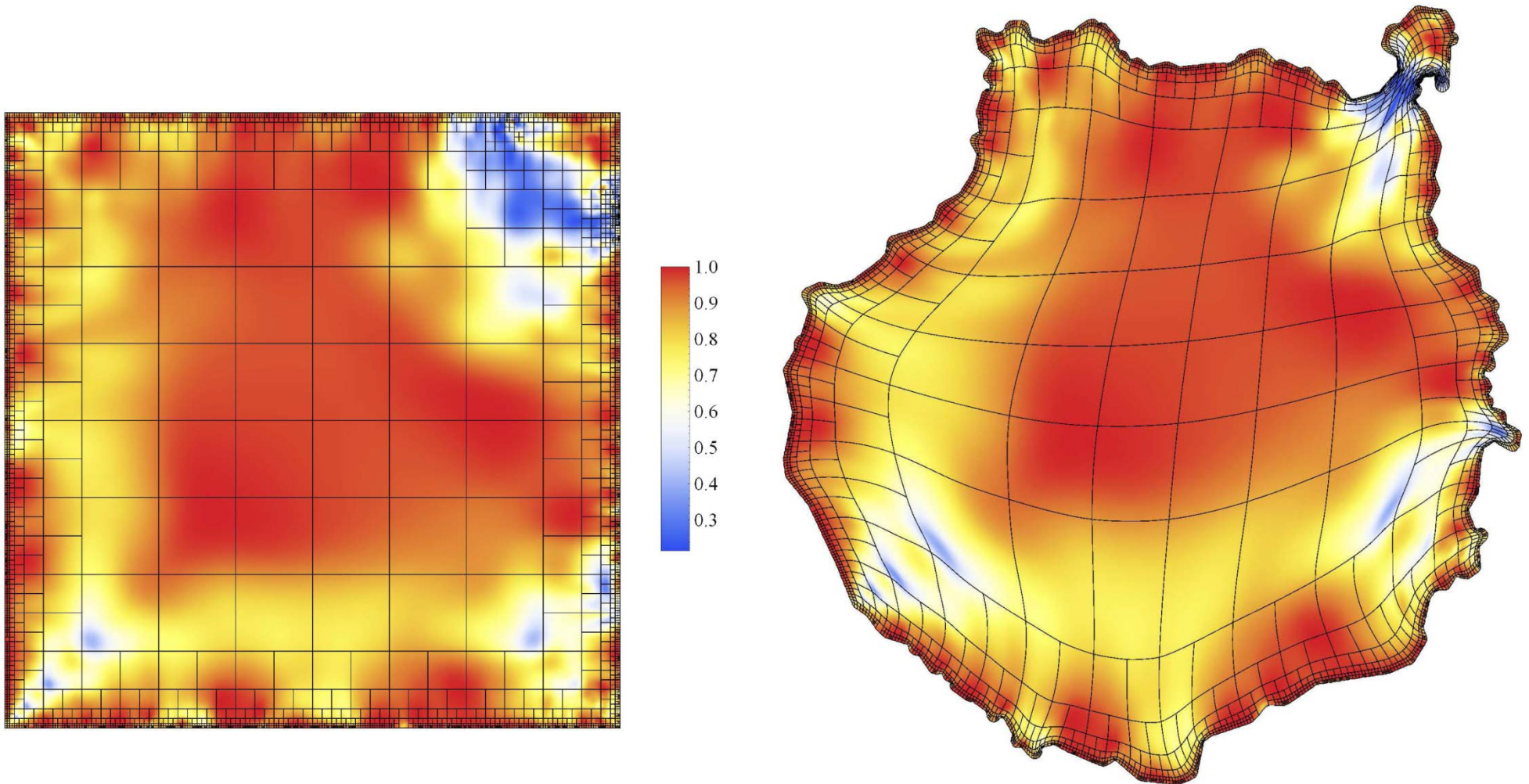
Spline parameterization of 2D geometries

Puzzle piece (Mean Ratio Jacobian)



Good quality parameterization for application of IGA

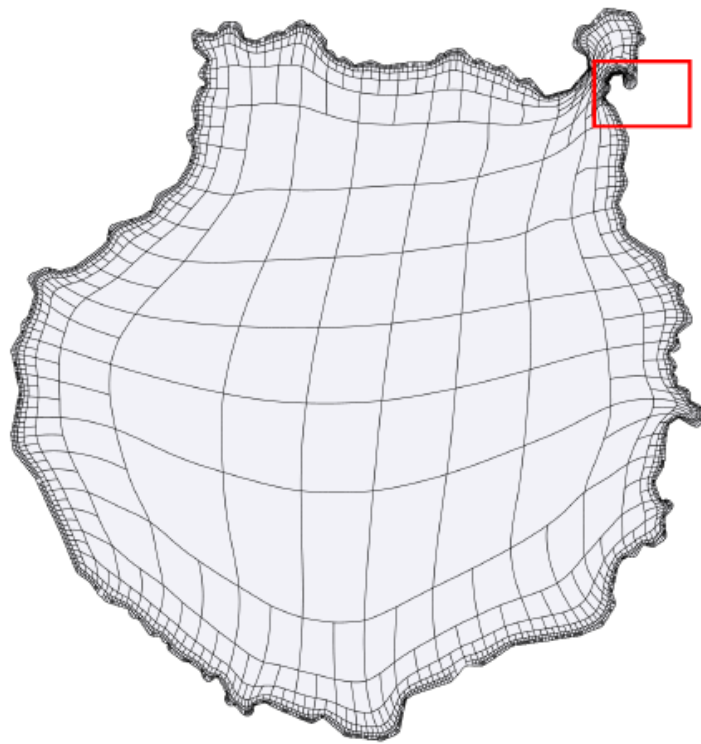
Mean ratio Jacobian as a quality metric



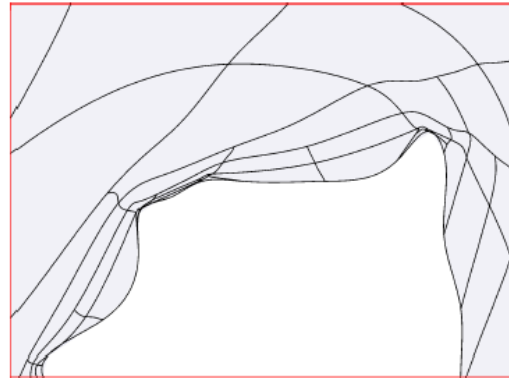
Spline parameterization of 2D geometries

Adaptive refinement to improve parameterization quality

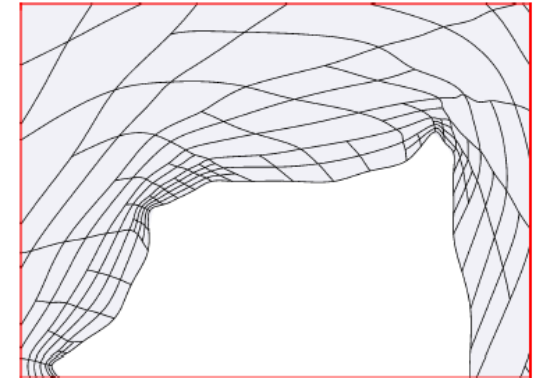
Low quality zone is refined and optimized again



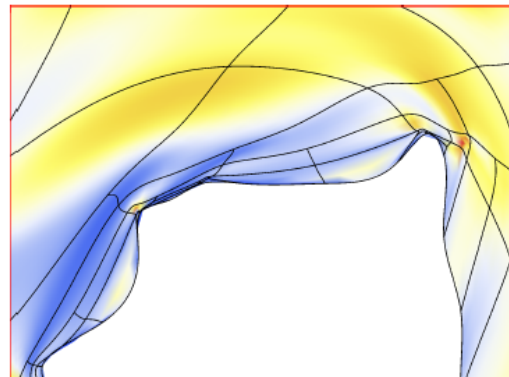
Isla de Gran Canaria



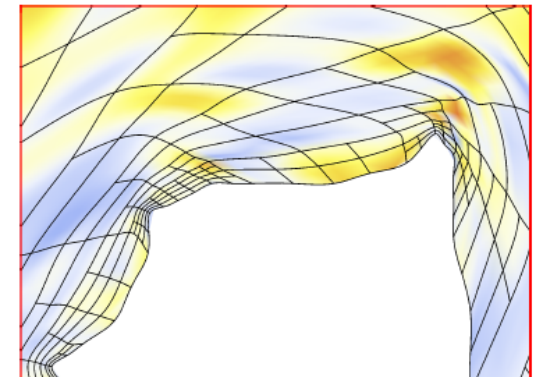
inicial T-spline



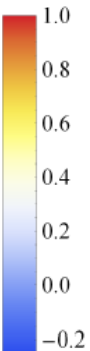
improved T-spline



Mean ratio Jacobian

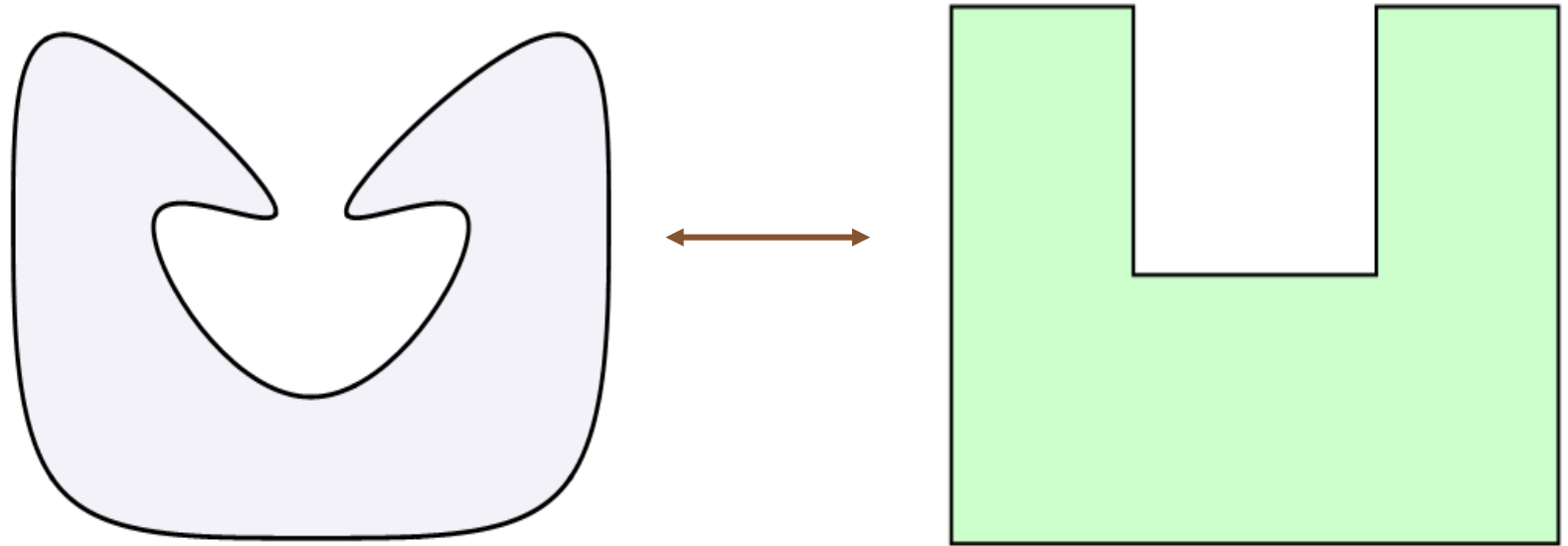


Mean ratio Jacobian



- What can we parameterize with a square? Something similar to a square and a little bit more.

We need more complex polycube-type parametric domain that fits better the geometry
[Li (2007), Liu (2014)]



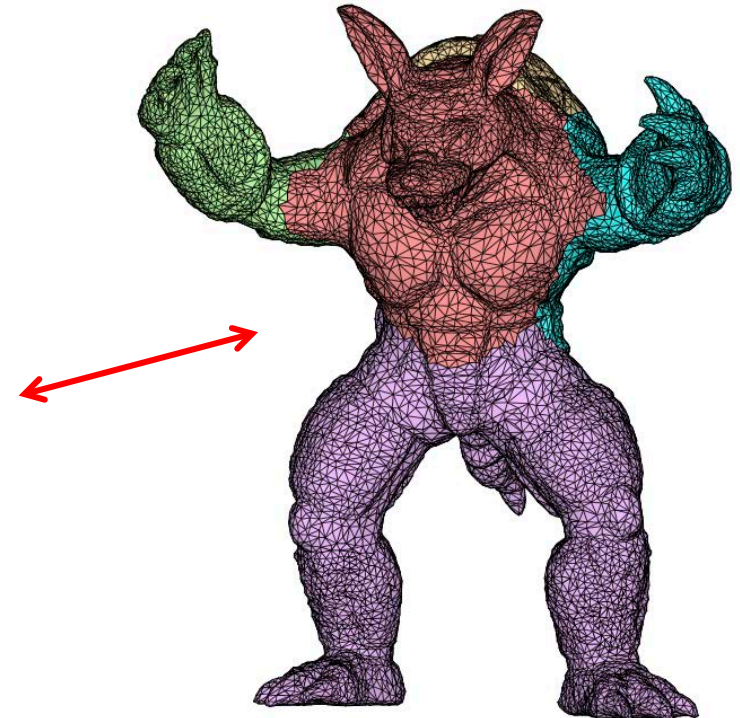
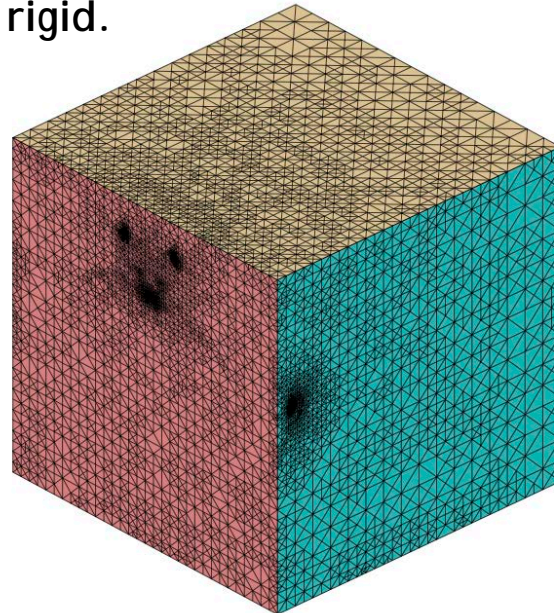
Spline parameterization of 2D and 3D geometries

Limitations and future research

All the above mentioned problems for 2D domain are aggravated considerably for 3D object:

- ❑ An appropriate selection of the edges may not exist
- ❑ More complex untangling and optimization procedure, the number of options increases a lot (O. Ushakova: Nondegeneracy tests for hexahedral cells, CMAME 2011)
- ❑ Jacobian close to zero along the edges of the object

Tensor product structure is too rigid.



Our results:

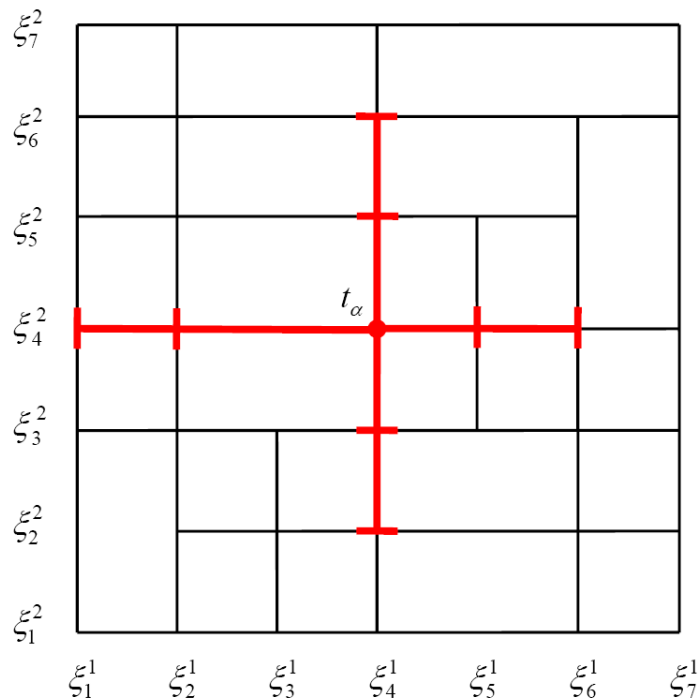
Polynomial spline spaces over hierarchical T-meshes

- A new strategy for constructing cubic spline space over arbitrary quadtree (2D) and octree (3D) T-meshes
- ❑ For a given T-mesh, it allows to obtain a set of cubic spline functions that span a space with nice properties: C^2 continuous, nested spaces, linear independence
- ❑ Simple rules for inferring local knot vectors to define spline blending functions
- ❑ Straightforward implementation in 2D and 3D

Spline spaces over T-meshes

T-splines, Sederberg (2003)

T-mesh and anchor t_a

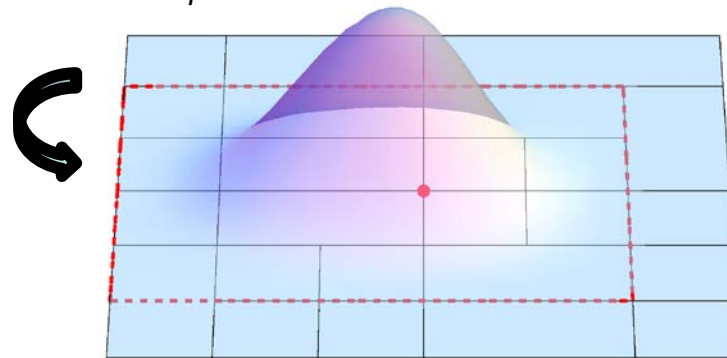


Knots associated to anchor t_a :

$$\Xi_\alpha^1 = \{\xi_1^1, \xi_2^1, \xi_4^1, \xi_5^1, \xi_6^1\} \quad \Xi_\alpha^2 = \{\xi_2^2, \xi_3^2, \xi_4^2, \xi_5^2, \xi_6^2\}$$

$$N_\alpha^1(\xi^1) \times N_\alpha^2(\xi^2) = B_\alpha(\xi^1, \xi^2) = N_\alpha^1(\xi^1) N_\alpha^2(\xi^2)$$

support of the T-spline



Bivariate Cubic T-spline Basis Function

Normalized rational T-splines

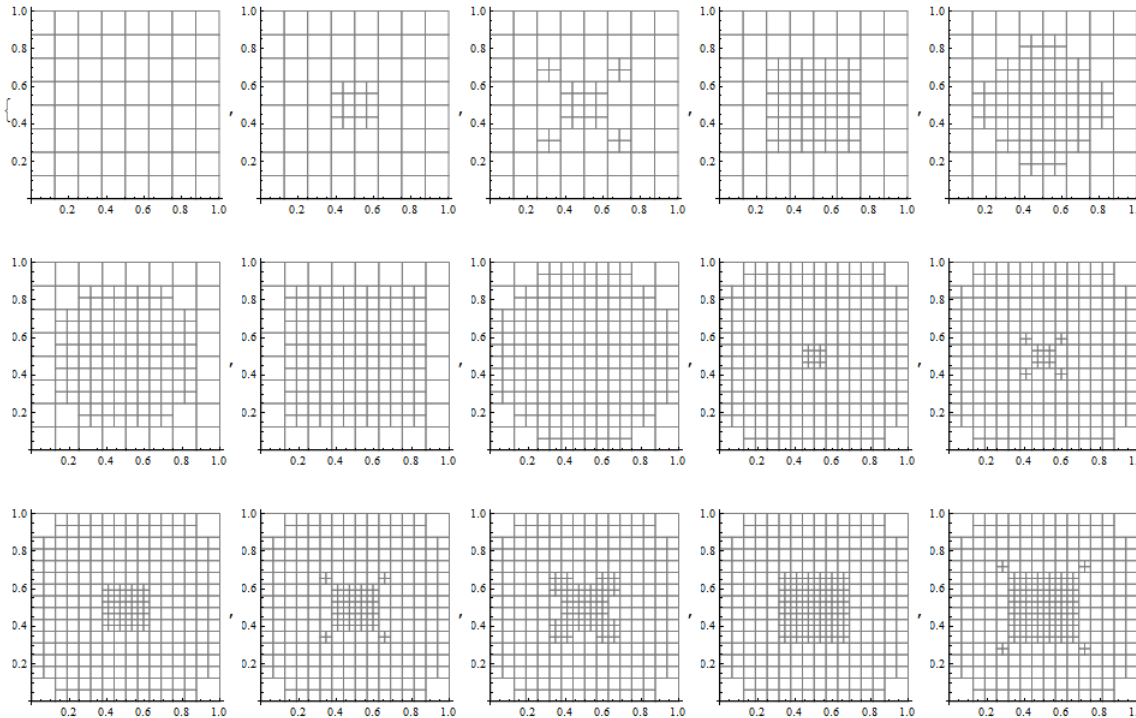
$$R_\alpha(\xi^1, \xi^2) = \frac{B_\alpha(\xi^1, \xi^2)}{\sum_{\beta \in A} B_\beta(\xi^1, \xi^2)}$$

Spline spaces over T-meshes

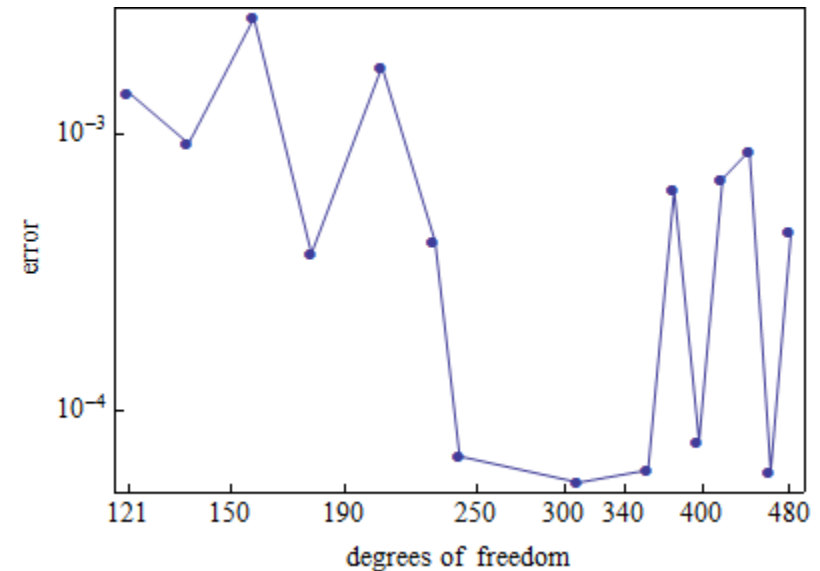
T-splines, Sederberg (2003). Approximation capability of rational basis functions.



Adaptive refinement for the approximation of smooth gaussian-type function $u(x,y) = e^{-(x^2+y^2)}$ via interpolation with rational T-splines



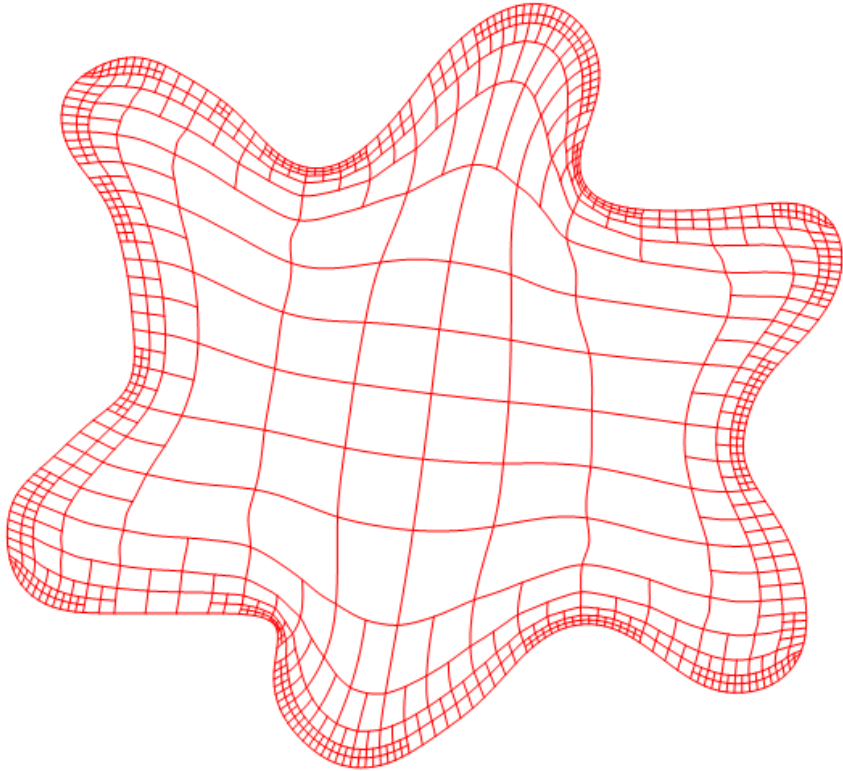
Error convergence en L^2 norm



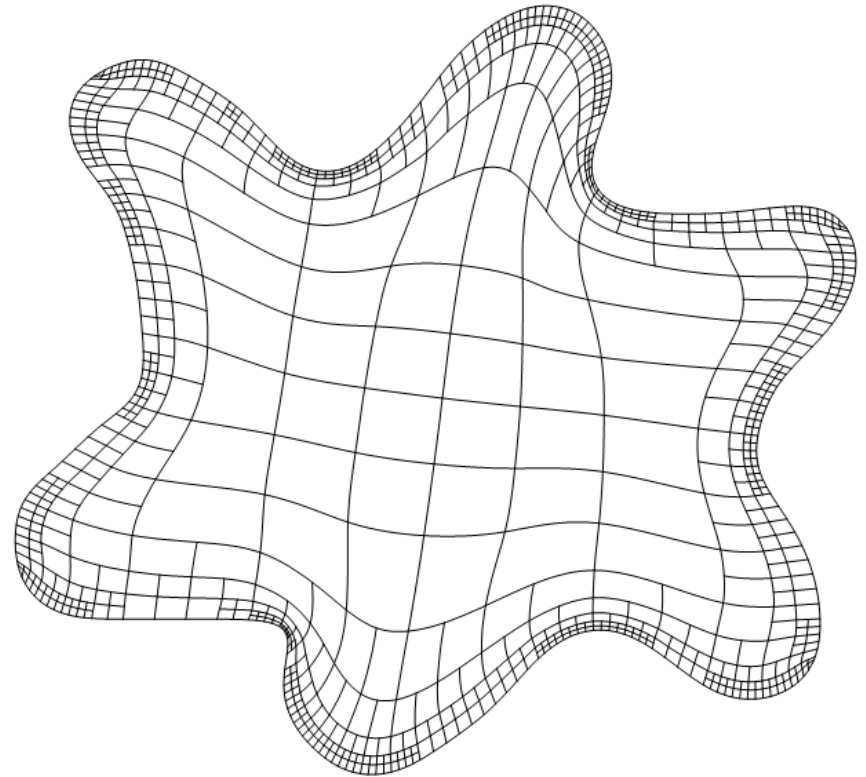
It is time to drop the “R” from NURBS, EWC, 2014
Les A. Piegl • Wayne Tiller • Khairan Rajab

Spline spaces over T-meshes

T-splines, Sederberg (2003). Approximation capability?



T-spline parameterization, Sederberg
(rational blending functions)

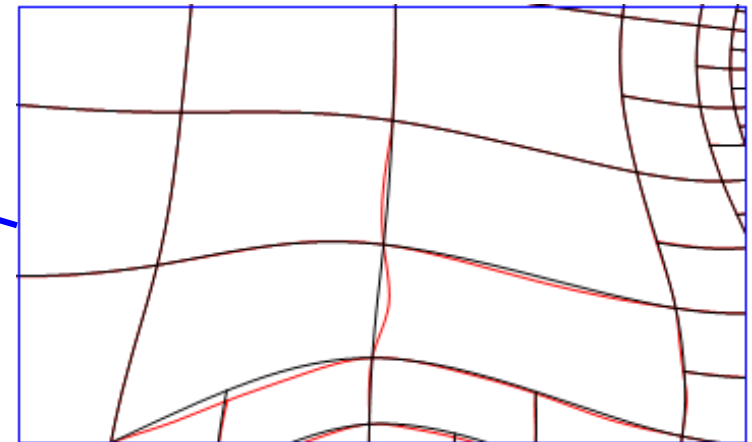
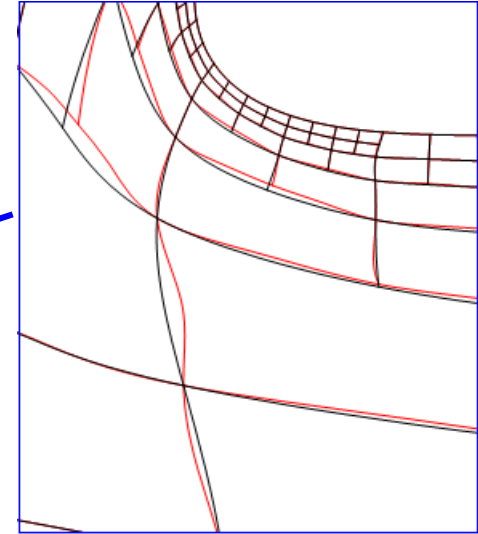
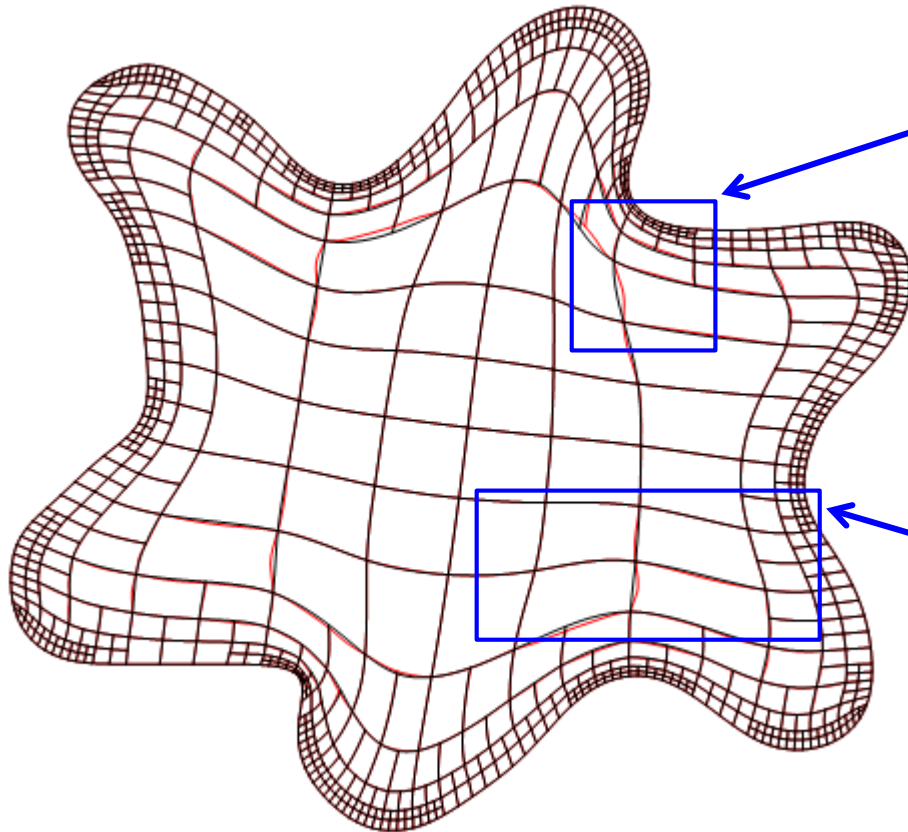


Spline parameterization with the new strategy
(polynomial blending functions)

Good quality parameterization for application of IGA

Comparison of rational T-spline parameterization with the new strategy

rational T-spline parameterization presents more oscillation

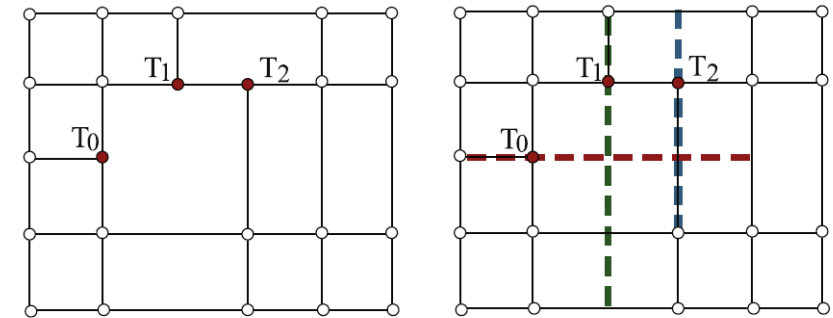


Spline spaces over T-meshes

Available strategies

Analysis suitable T-spline, Scott (2011)

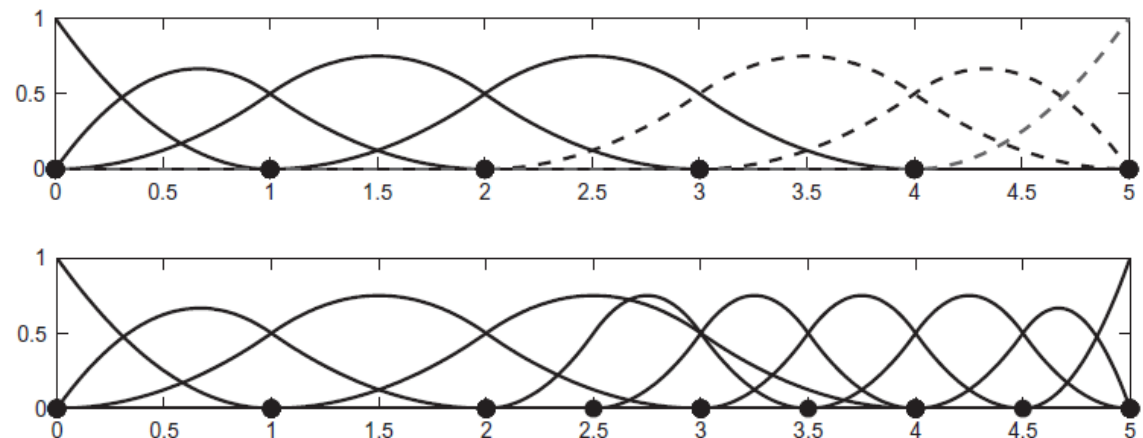
- nested spaces
- linear independence
- optimal support overlapping
- topological restriction for the T-mesh
difficult to implement => Does not exist in 3D yet!



(a) Extensions that cross.

Hierarchical refinement for IGA, Vuong (2011), Schillinger (2012), Bornemann(2012)

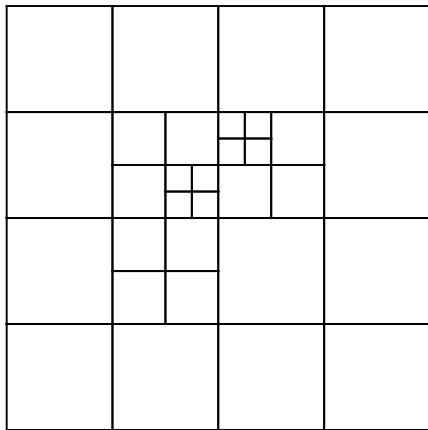
- nested spaces
- linear independence
- relatively easy implementation
- excessive support overlapping
- impossibility to define spline space over an arbitrary mesh



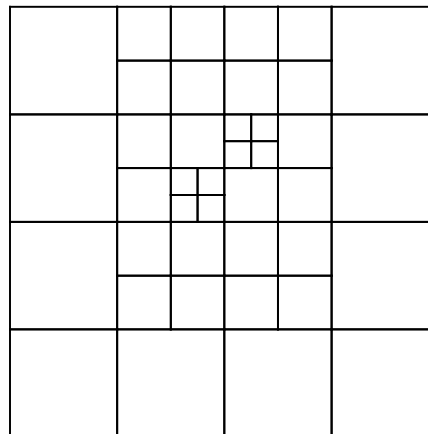
Our objective

- simple and easy implementation
- nice properties: linear independence and nested spaces

We work with 0-balanced quadtree and octree T-meshes



unbalanced T-mesh



0-balanced T-mesh

Steps to define spline basis functions over a given T-mesh:

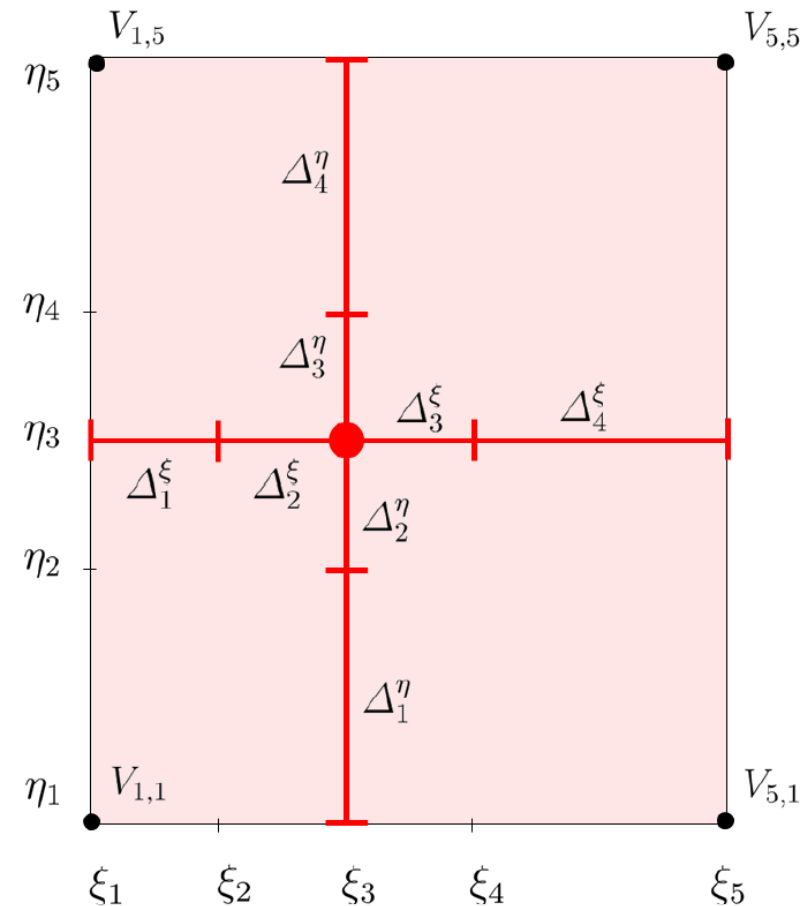
- Mesh pretreatment: 0-balancing
- Inferring local knot vectors
- Modification of local knot vectors

Condition 1: Local knot vectors verify:

$$\Delta_1^j \geq \Delta_2^j = \Delta_3^j \leq \Delta_4^j$$

Condition 2: The frame of the function support should be situated over the mesh skeleton:

$$\text{frm}(\text{supp } N_\alpha) \in \text{skt}(T)$$

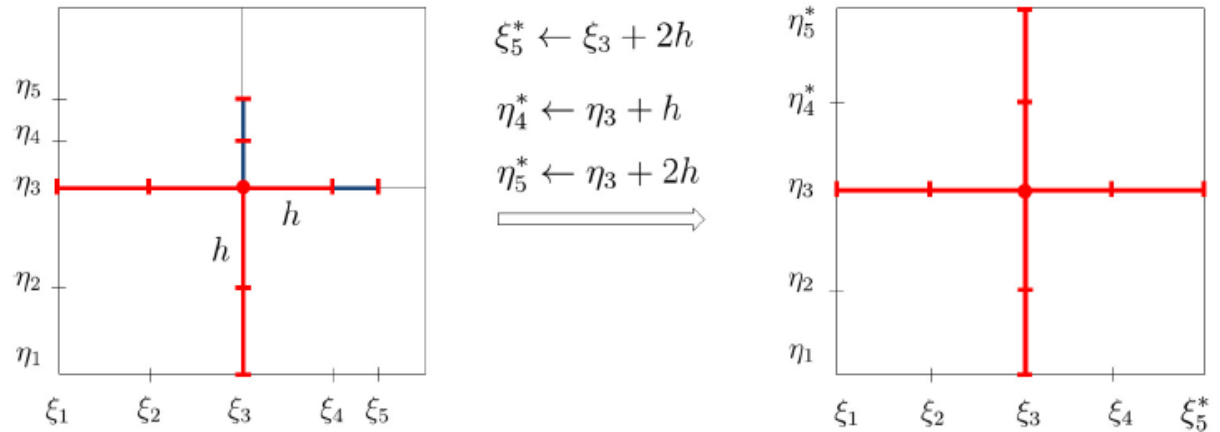


Spline spaces over T-meshes

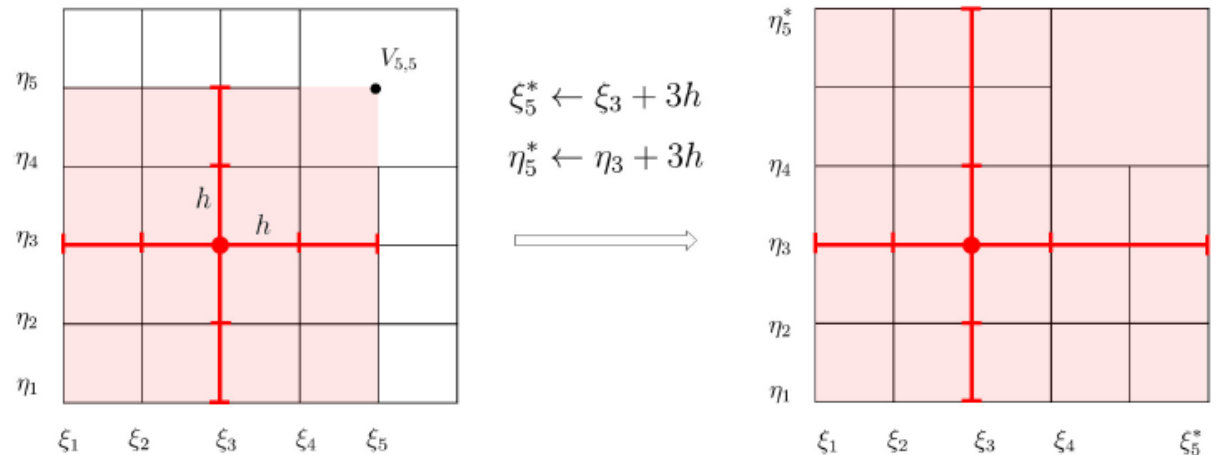
Definition of polynomial spline functions. Extension rules in 2D

Support modification
to fulfill the imposed conditions

Simple extension rules:
duplicate some knot intervals



(a) Condition 1 is not satisfied because $\Delta_4^\xi < \Delta_3^\xi$ and $\Delta_3^\eta < \Delta_2^\eta$.



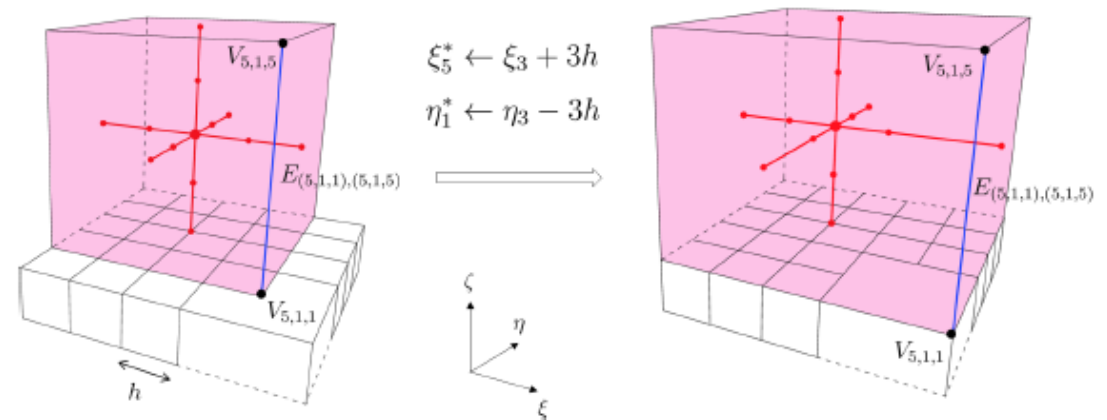
(b) Condition 2 is not satisfied because $V_{5,5} \notin \text{skt}(T)$.

Spline spaces over T-meshes

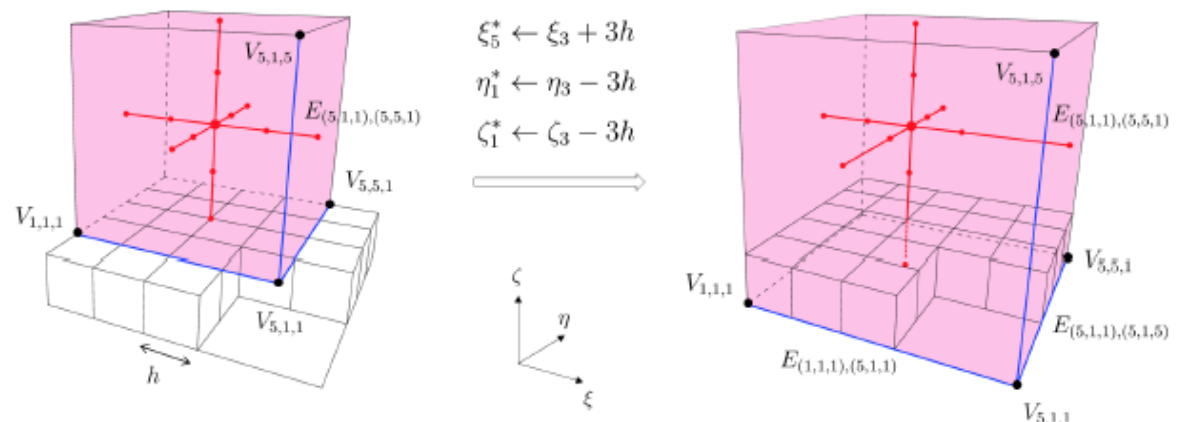
Definition of polynomial spline functions. Extension rules in 3D

Support modification
to fulfill the imposed conditions

Simple extension rules in 3D as well !



(a) Only one edge violating Condition 2. The node $V_{5,1,1}$ is sited in the center of the face of size $2h$. The support is extended in two directions.



(b) Three edges violating Condition 2. The node $V_{5,1,1}$ is sited in the center of the cell of size $2h$. The support is extended in three directions.

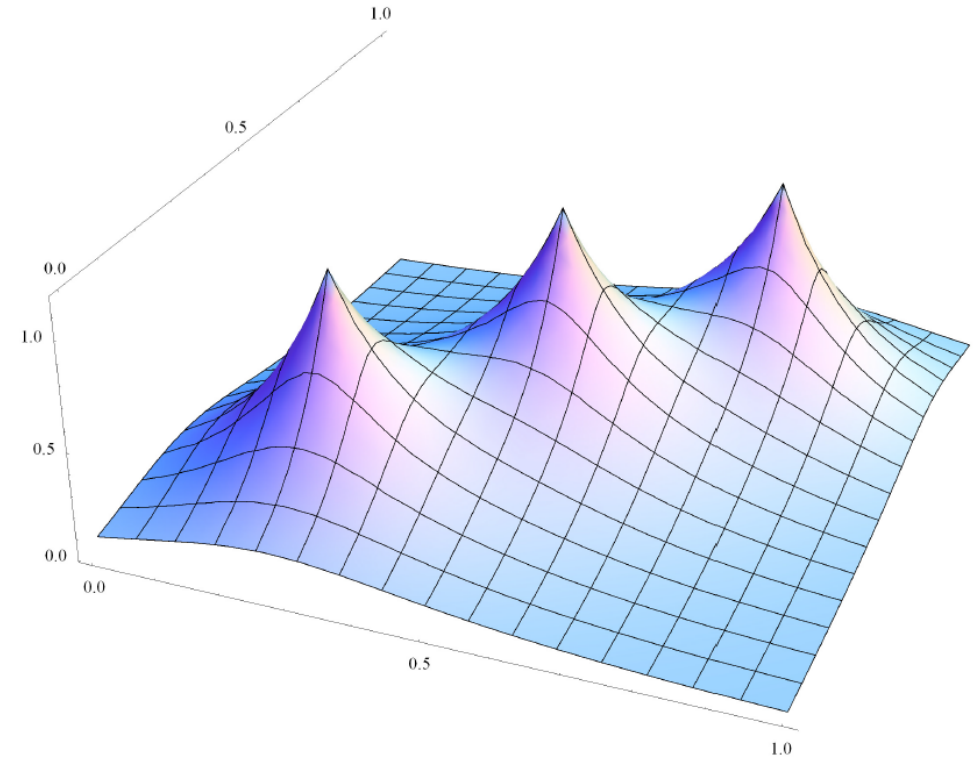
Isogeometric analysis on 2D domain

Poisson problem with Dirichlet boundary condition

$$\begin{aligned} -\Delta u &= f && \text{in } \Omega, \\ u &= g && \text{on } \partial\Omega. \end{aligned}$$

Analytical solution:

$$\begin{aligned} u(x, y) = & \exp\left(-7\sqrt{(x-0.5)^2 + (y-0.5)^2}\right) + \\ & + \exp\left(-7\sqrt{(x-0.25)^2 + (y-0.25)^2}\right) + \\ & + \exp\left(-7\sqrt{(x-0.75)^2 + (y-0.75)^2}\right). \end{aligned}$$

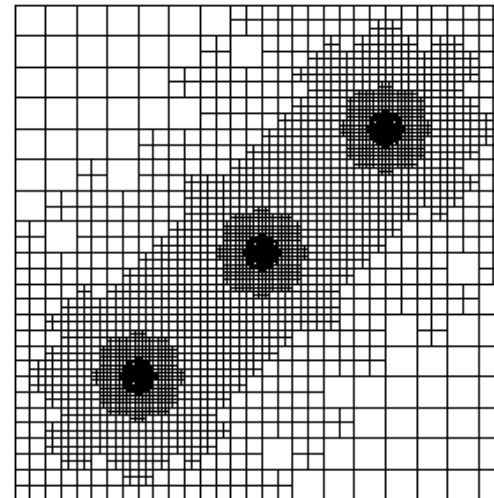
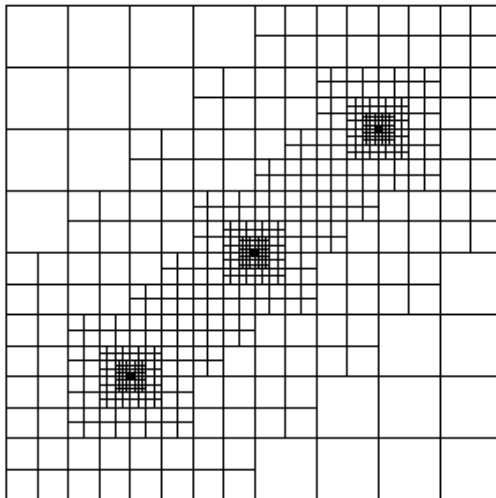
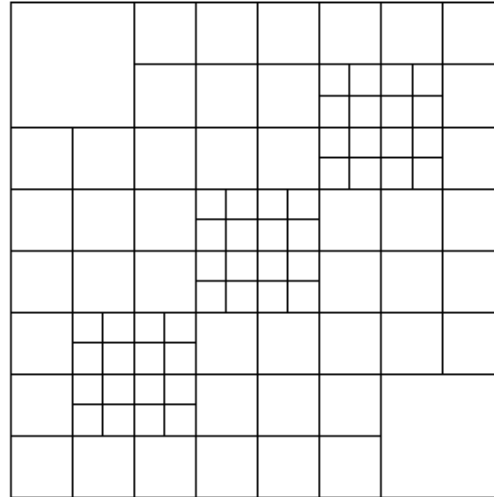
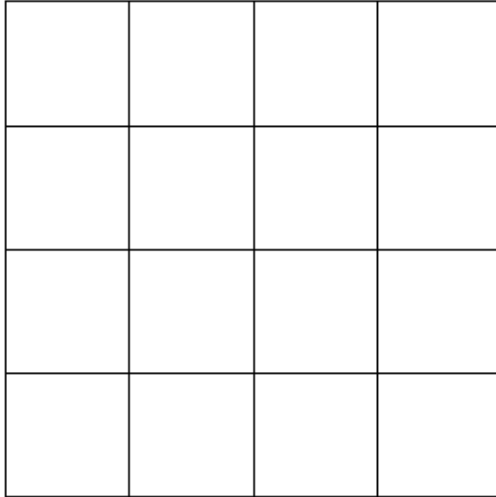


Residual-type error indicator:

$$\eta(\Omega_e)^2 = \int_{\Omega_e} h^2 (f + \Delta u_h)^2 \, d\Omega$$

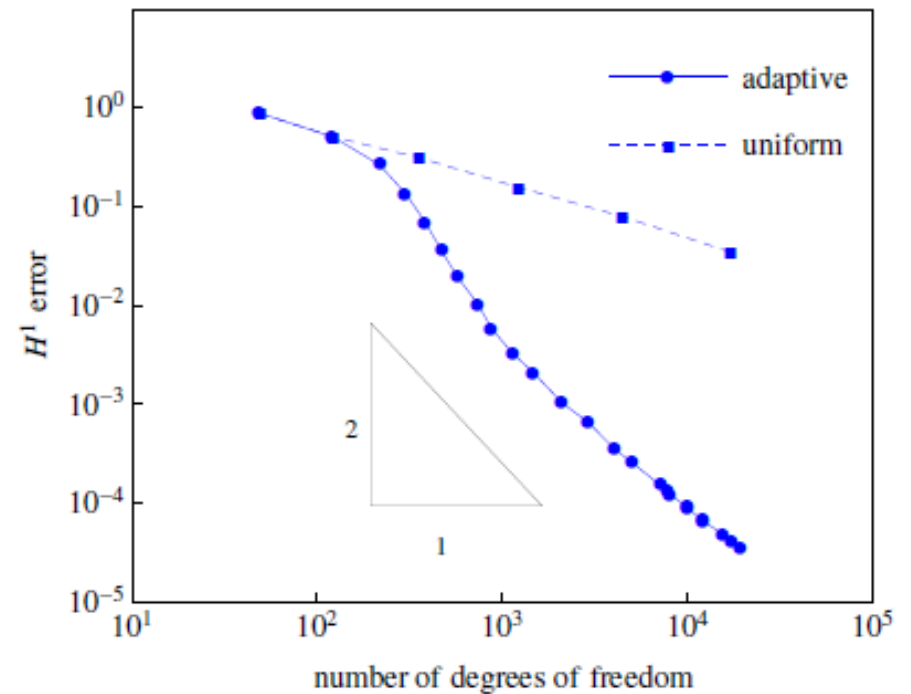
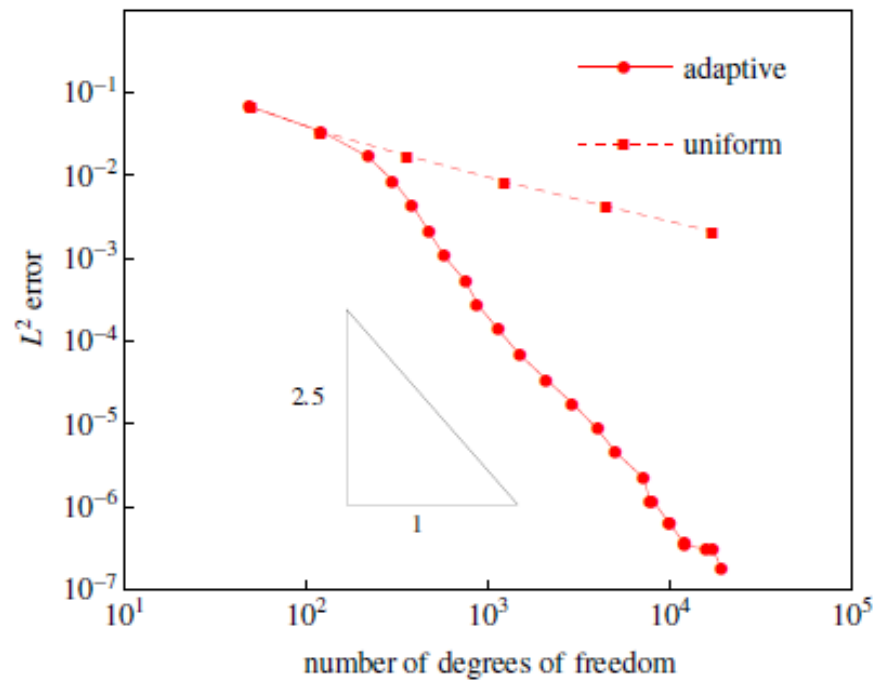
Adaptive refinement for 2D Poisson problem

Error indicator : $\eta(\Omega_e)^2 = \int_{\Omega_e} h^2 (f + \Delta u_h)^2 d\Omega$

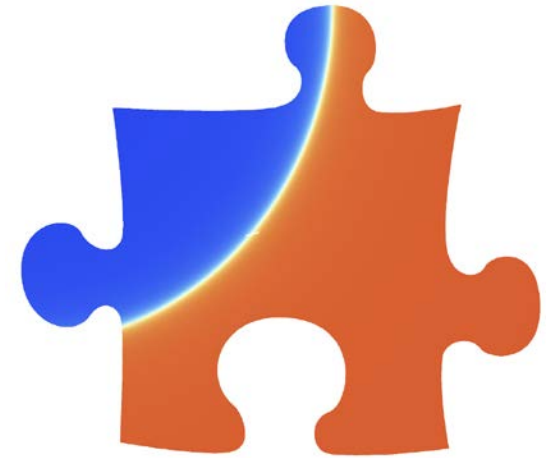


Adaptive refinement for 2D Poisson problem

Monotonous convergence: L^2 -norm and H^1 -seminorm error



$$\begin{aligned} -\Delta u &= f && \text{in } \Omega, \\ u &= g && \text{on } \partial\Omega. \end{aligned}$$



- Analytical solution: Steep wave front given by

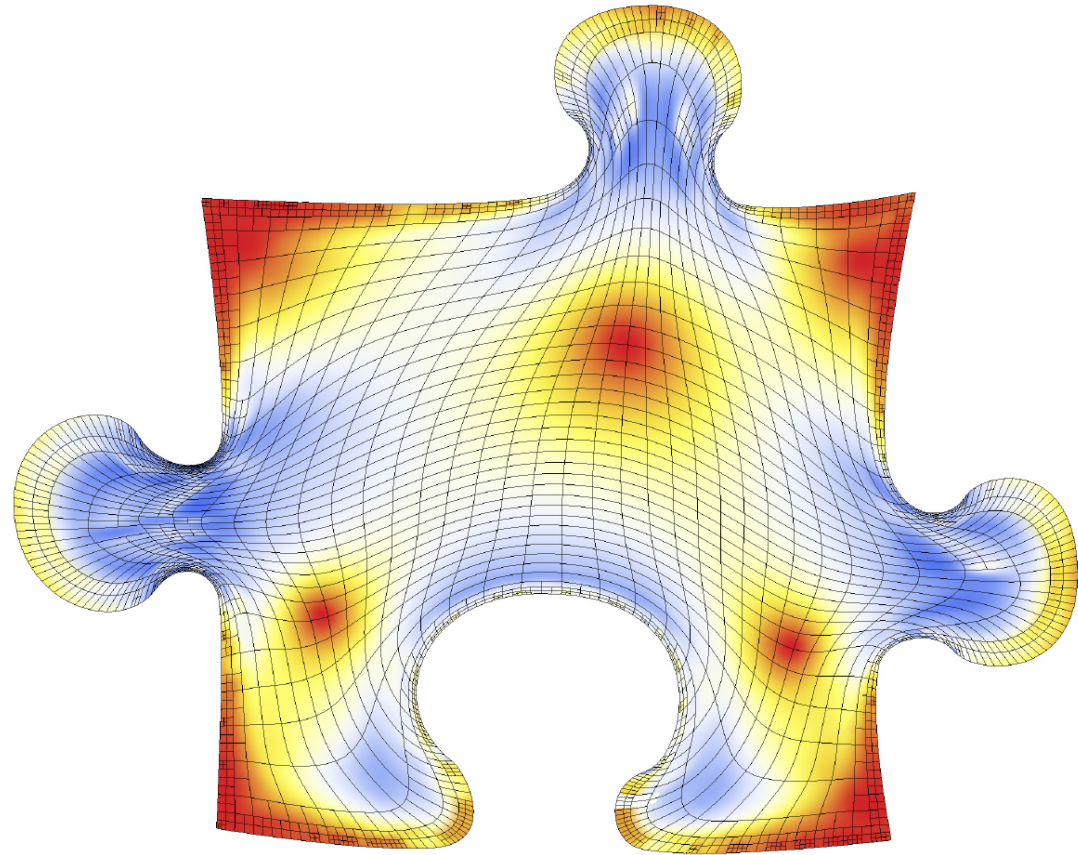
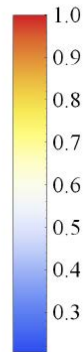
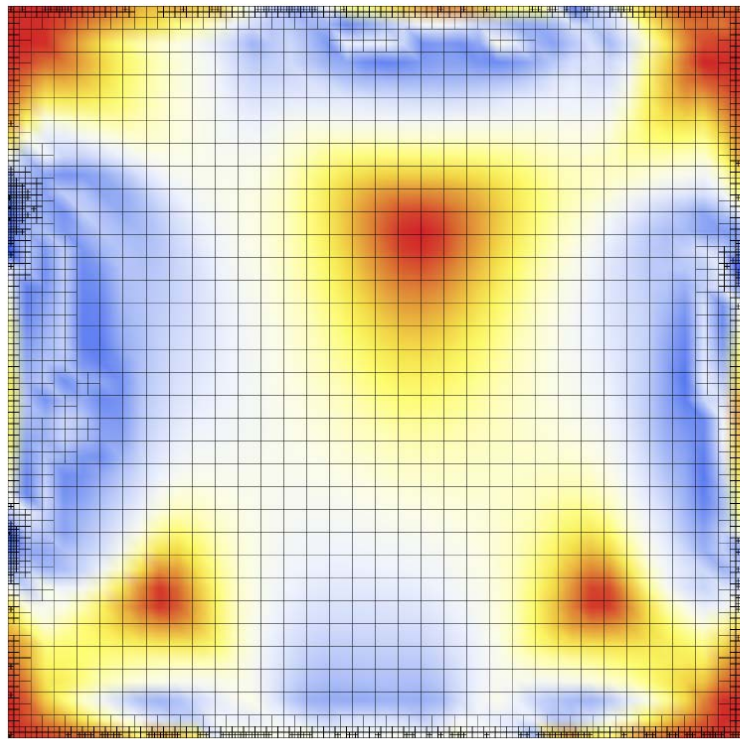
$$u(r) = \arctan(\alpha(r - r_0)), \quad \text{where } r = \sqrt{(x - x_c)^2 + (y - y_c)^2}, \\ (x_c, y_c) = (0, 0), \alpha = 200 \text{ and } r_0 = 0.6$$

- Adaptive strategy: Residual-type error indicator:

$$\eta(\Omega_e)^2 = \int_{\Omega_e} h^2 (f + \Delta u_h)^2 \, d\Omega$$

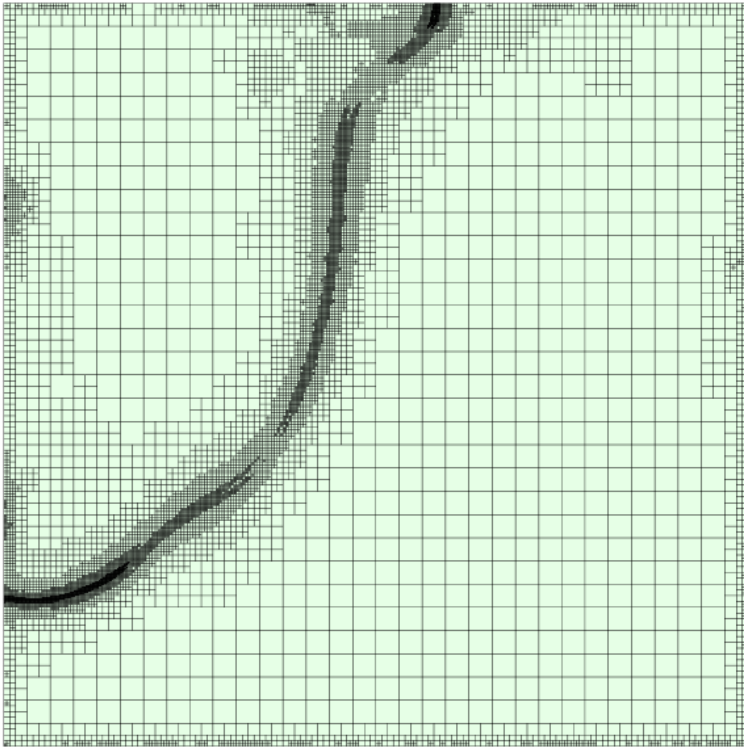
Isogeometric analysis for 2D Poisson problem

Parameterization of computational domain

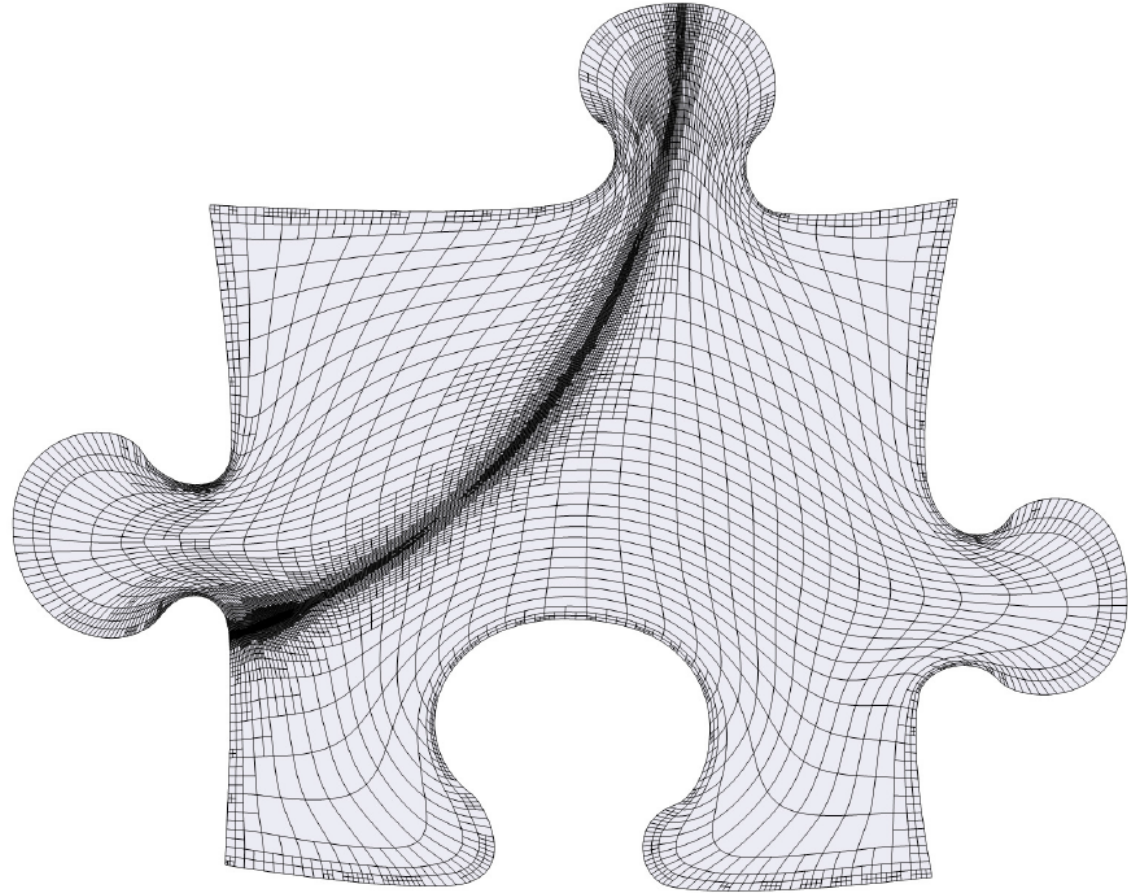


Adaptive refinement for 2D Poisson problem

Adaptive T-mesh after 13 refinement steps



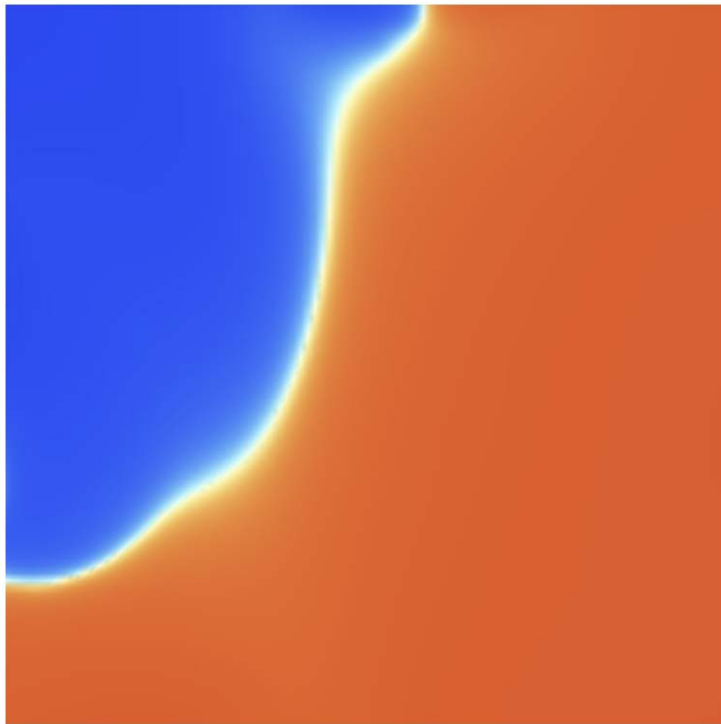
final refinement in the parametric domain



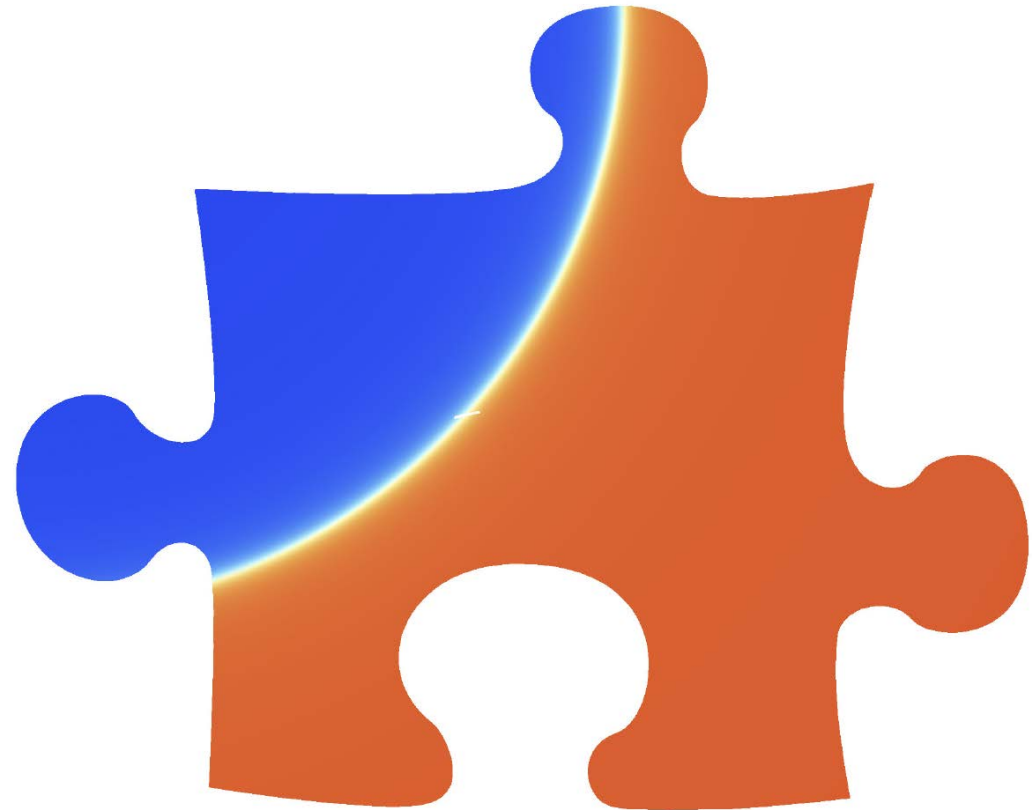
final refinement in the physical domain

Adaptive refinement for 2D Poisson problem

Numerical solution



numerical solution in the parametric domain



numerical solution in the physical domain

Adaptive refinement for 2D Poisson problem

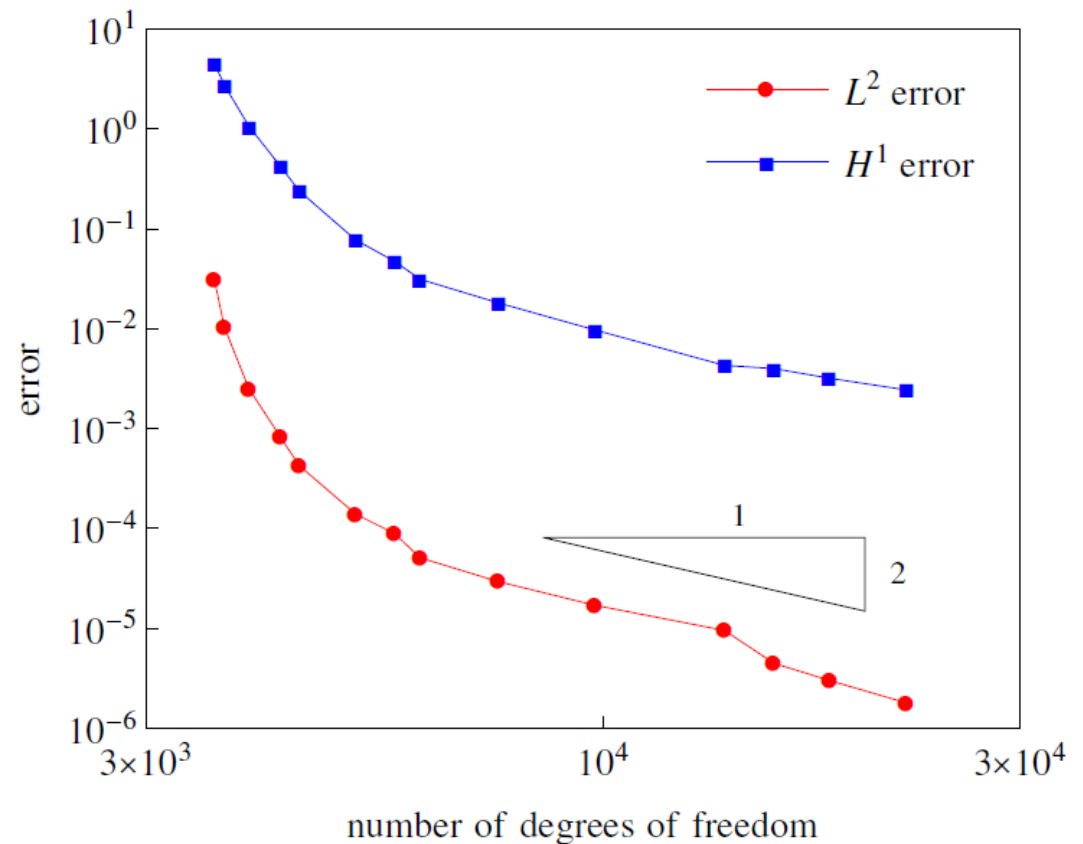
Convergence behavior

Convergence of L^2 -norm and H^1 -seminorm error

expected order for 2D problem:

L^2 -norm: 2

H^1 -seminorm: 3/2



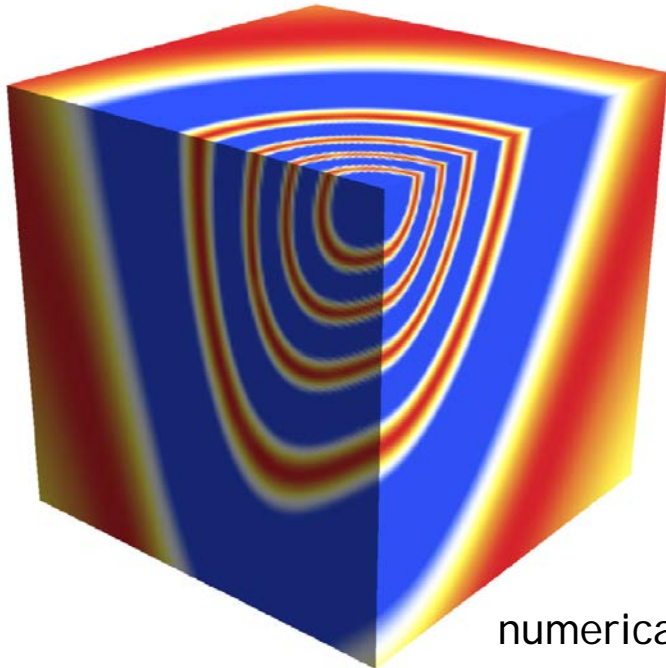
Isogeometric analysis for 3D Poisson problem

Computational domain: sphere portion

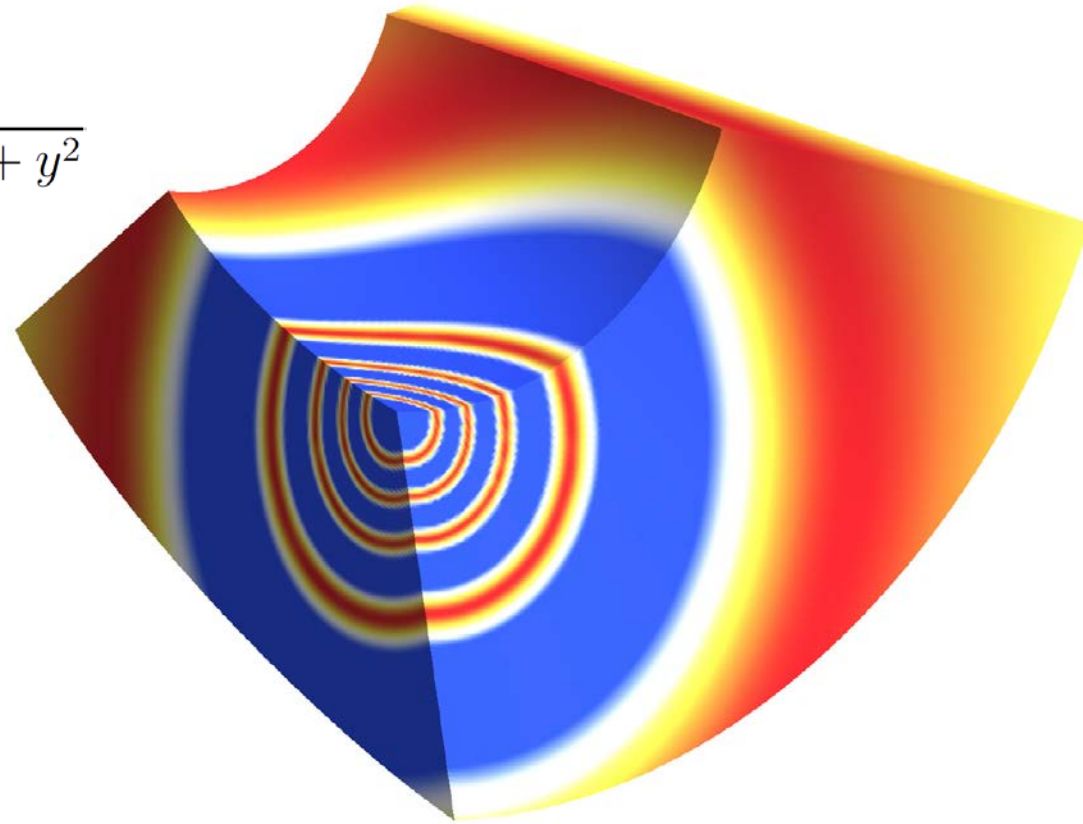
Analytical solution:

$$u(r) = \sin\left(\frac{1}{\alpha + r}\right), \quad \text{where } r = \sqrt{x^2 + y^2}$$

oscillation parameter $\alpha = 1/10\pi$



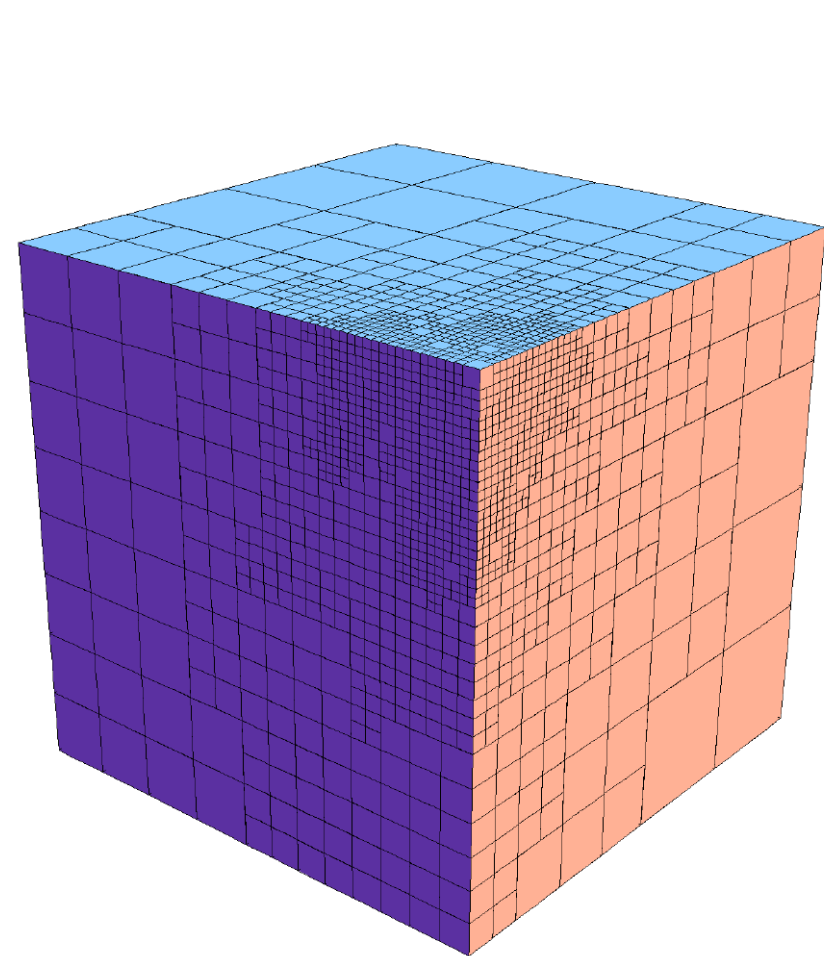
numerical solution in the parametric domain



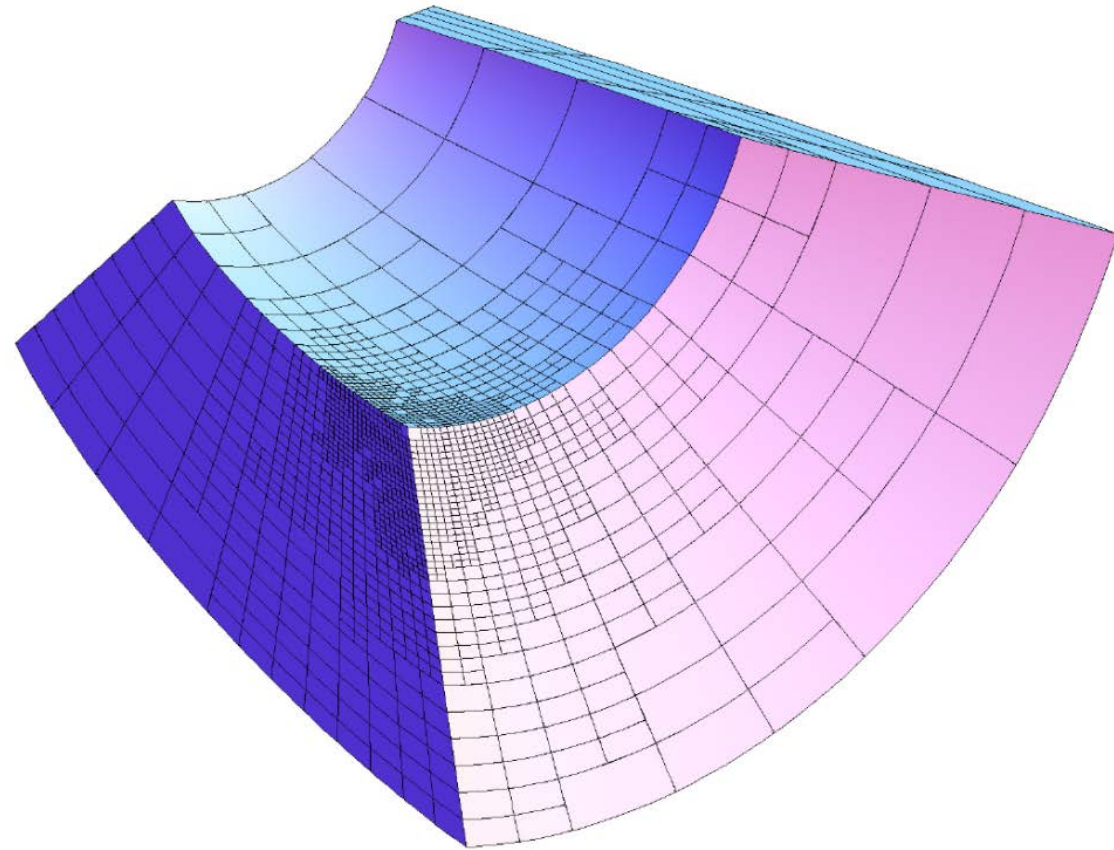
numerical solution in the physical domain

Adaptive refinement for 3D Poisson problem

Adaptive refinement



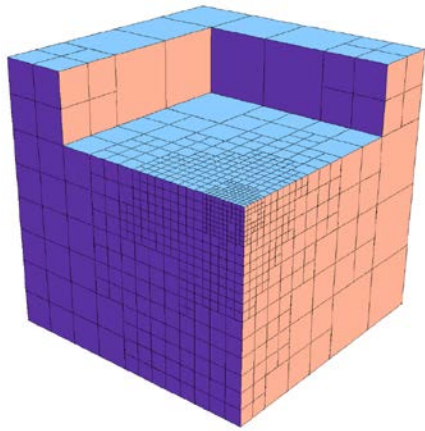
final refinement in the parametric domain



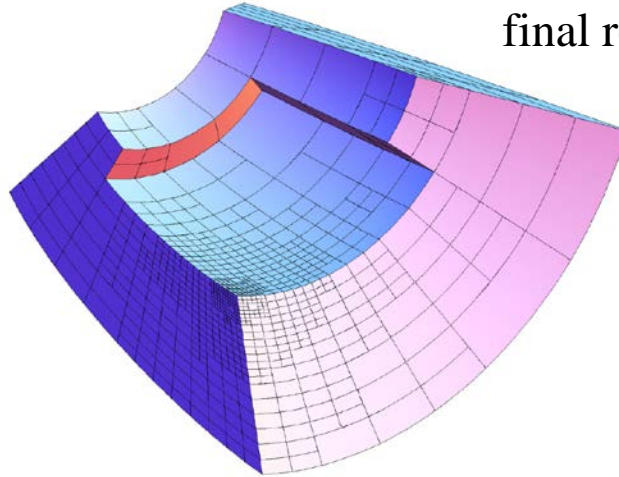
final refinement in the physical domain

Adaptive refinement for 3D Poisson problem

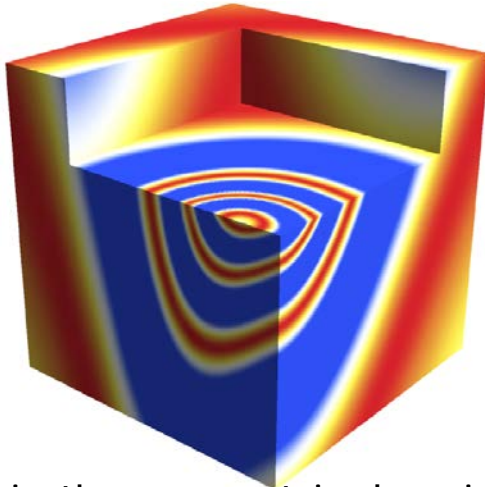
A section of the parametric and physical domain



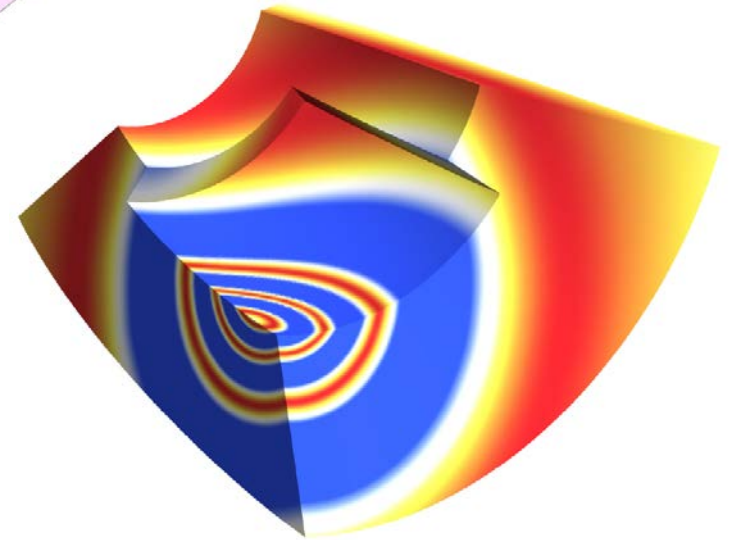
final refinement in the parametric domain



final refinement in the physical domain



numerical solution in the parametric domain



numerical solution in the physical domain

Adaptive refinement for 3D Poisson problem

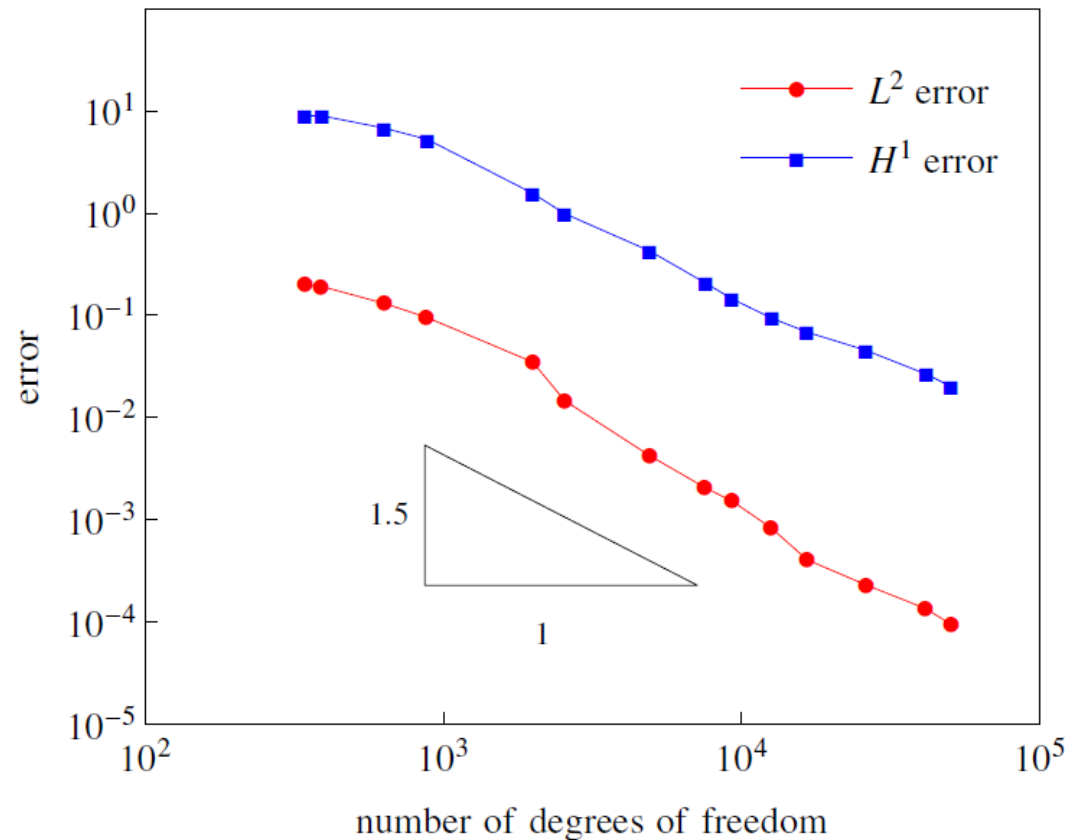
Convergence behavior

Convergence of L^2 -norm and H^1 -seminorm error

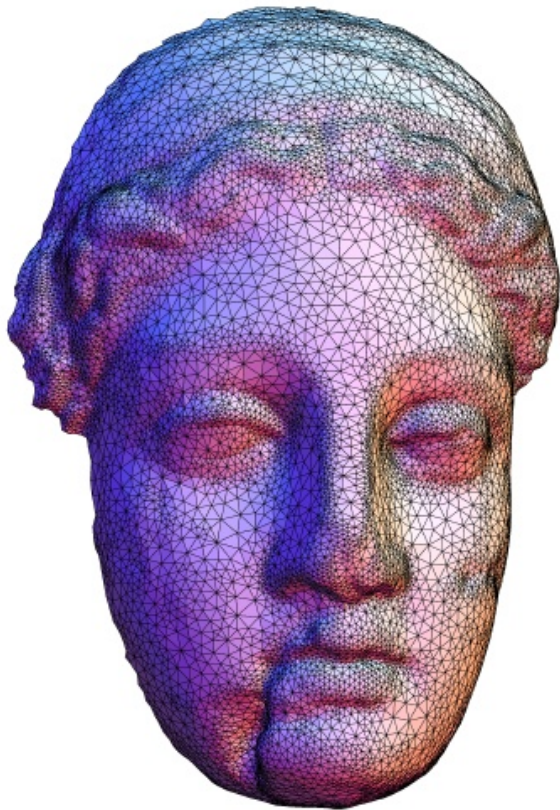
expected order for 3D problem:

L^2 -norm: $3/2$

H^1 -seminorm: 1



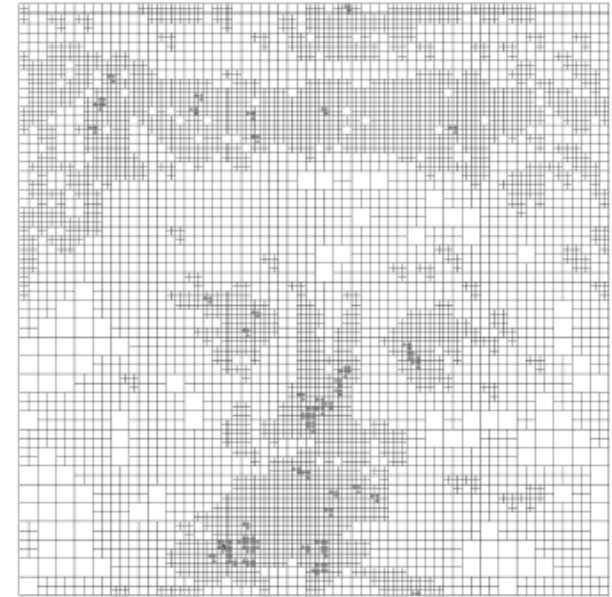
Spline representation of the surface from its triangulation



input triangulation



spline representation



parametric T-mesh

- Excessive overlapping of function support in some situations

Possible improvement of the locality of function supports

- Rigorous proof of the properties of the spaces defined with our strategy:

- linear independence
- nestedness of the spaces

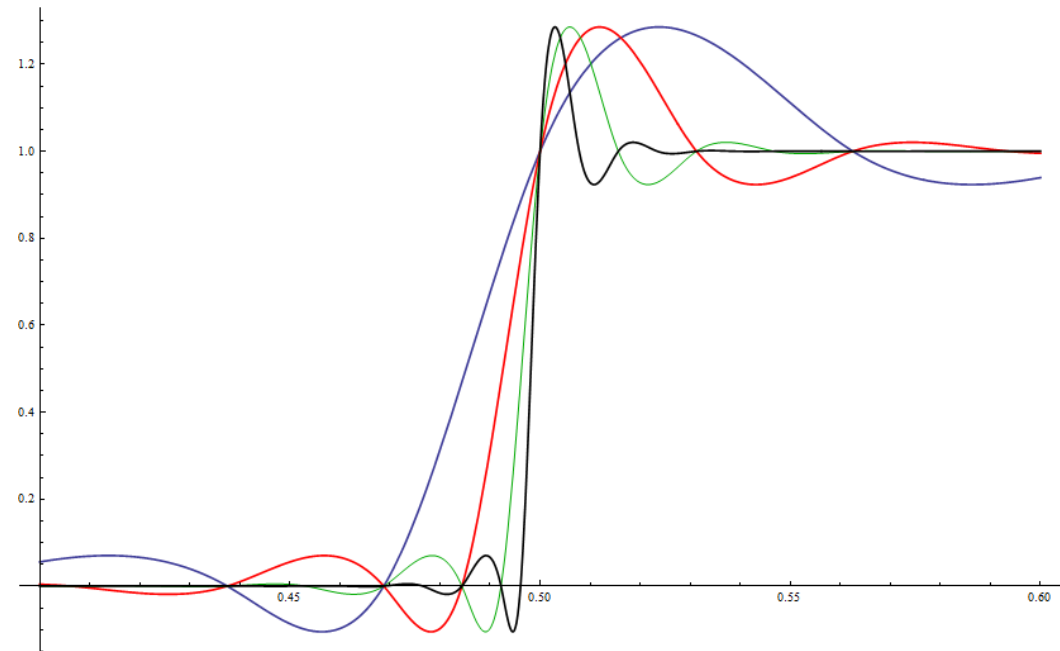
And another interesting question.

Is IGA always better than FEM ?

C^2 is always better than C^0 ?

It seems to be a lot better for approximation of very smooth function.

And not so smooth ??? Should be studied.



Spline approximation of a steep function

GRACIAS !