

Analysis-suitable parameterization for isogeometric analysis

Ye Ji (纪野)

Department of Applied Mathematics

Delft University of Technology, The Netherlands



Outline

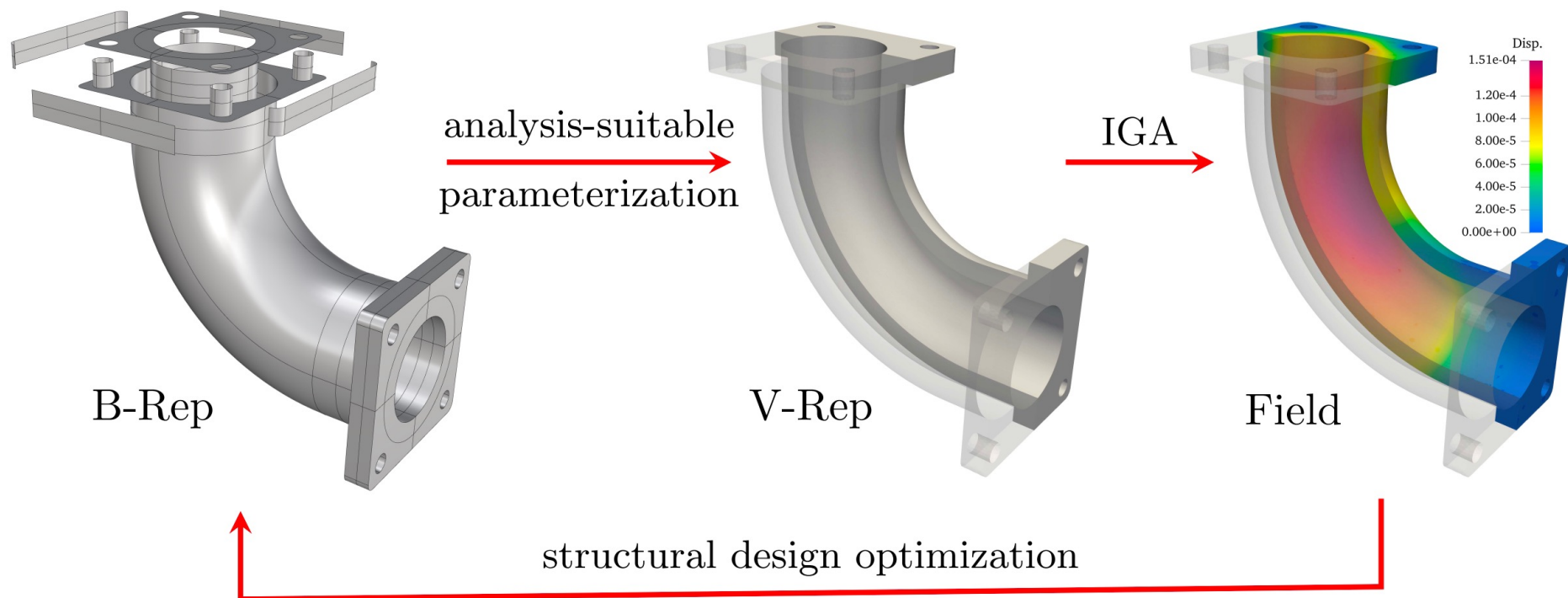
1. Background and motivation
2. Overview of the algorithms
3. Applications
4. Conclusions and outlook

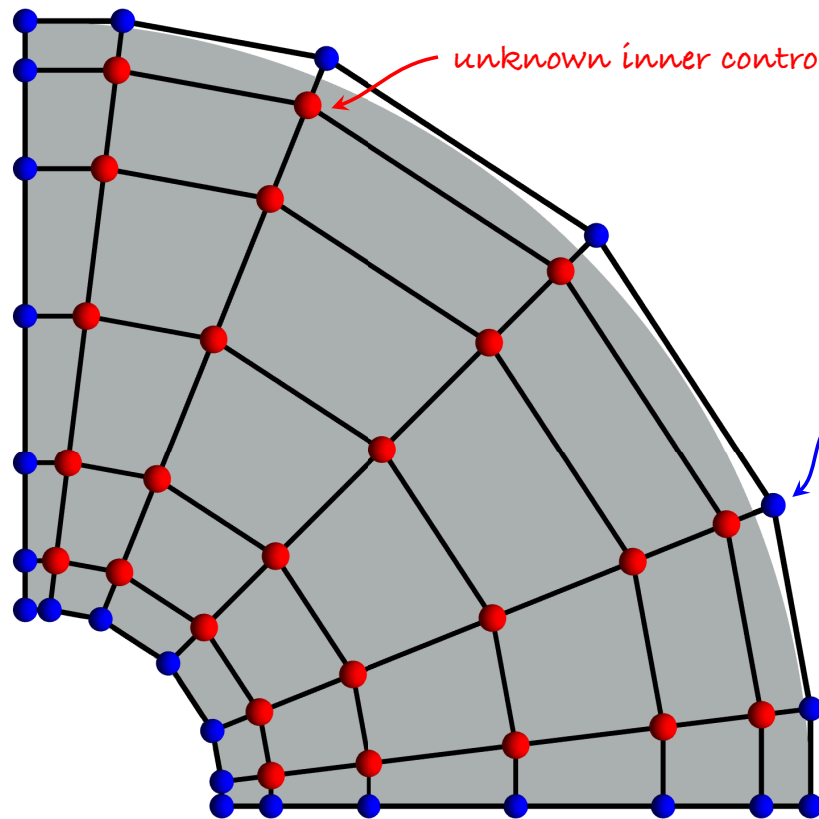


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- CAD models are usually represented by **boundary representation** (B-Rep);
- However, IGA requires an internal **spline-based parameterization** (V-Rep).





$$x(\xi) = \underbrace{\sum_{i \in \mathcal{I}_I} \mathbf{P}_i R_i(\xi)}_{\text{unknown}} + \underbrace{\sum_{j \in \mathcal{I}_B} \mathbf{P}_j R_j(\xi)}_{\text{known}}.$$

known boundary control points $\mathbf{P}_j$

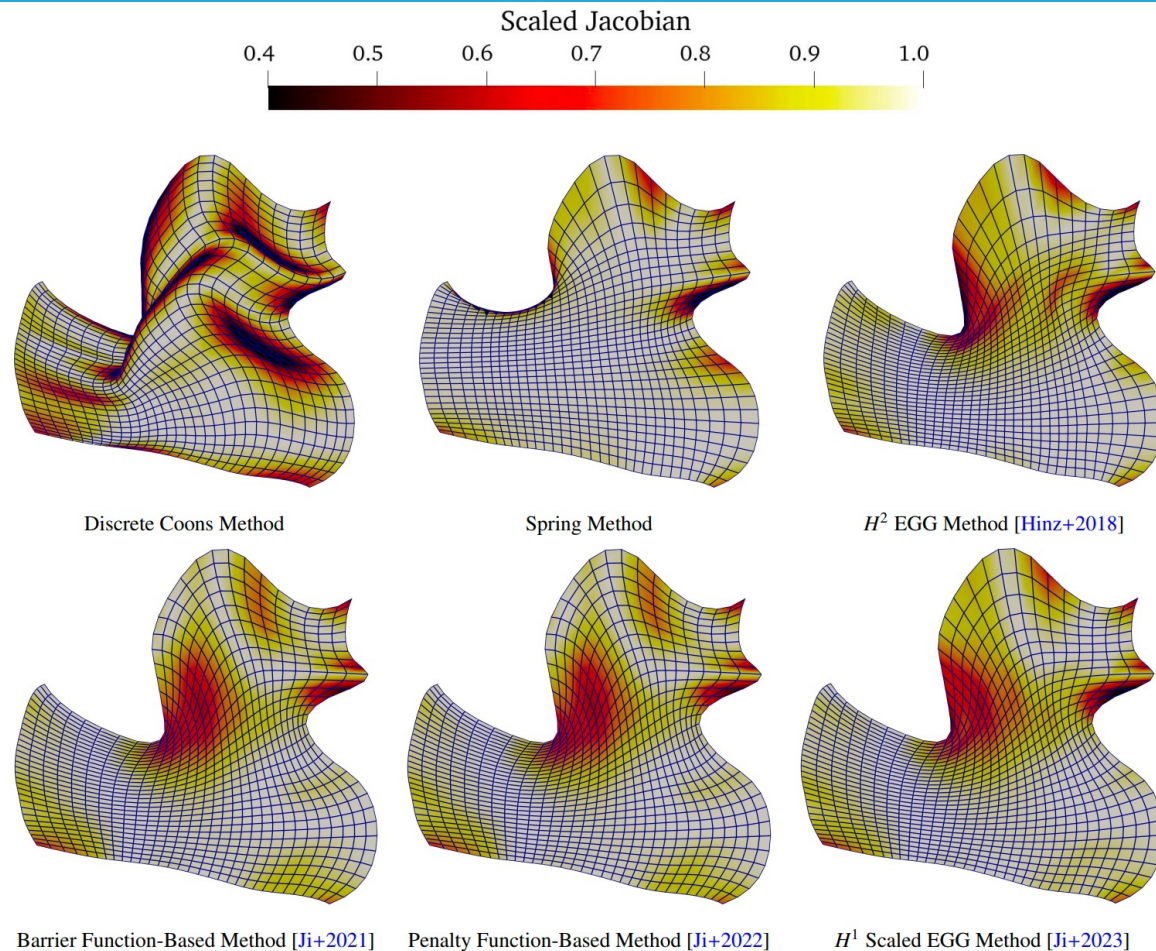
GOAL: to construct **unknown inner control points $\mathbf{P}_i$** (or basis functions R_i) such that x ensures **bijectivity** and exhibits **optimal orthogonality** and **uniformity**.

➤ **Parameterization quality** significantly affects **downstream analysis!**



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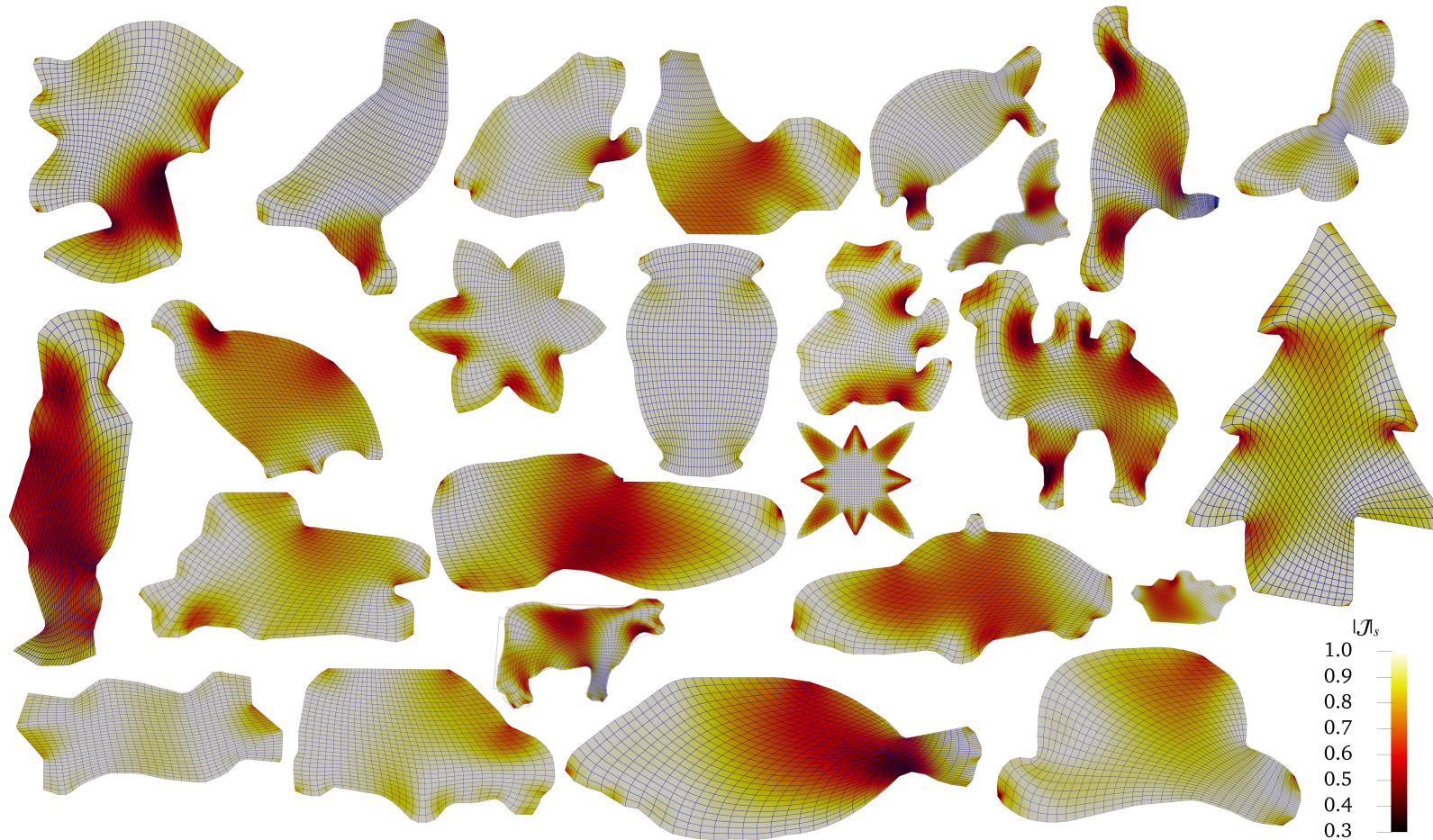
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- Algebraic methods:
 - Coons patch
 - Spring patch
- Optimization-based methods:
 - Barrier-function-based ¹
 - Penalty-function-based ²
- PDE-based methods:
 - Elliptic grid generation ³
 - Improved EGG ⁴

1. Ji, Y. et al. (2021). JCAM, 396, 113615.
2. Ji, Y. et al. (2022). CAGD, 94, 102081.
3. Hinz, J. et al. (2018). CAGD, 65, 48-75.
4. Ji, Y. et al. (2023). CAGD, 102, 102191.

Planar Parameterization Test Dataset (977 models)

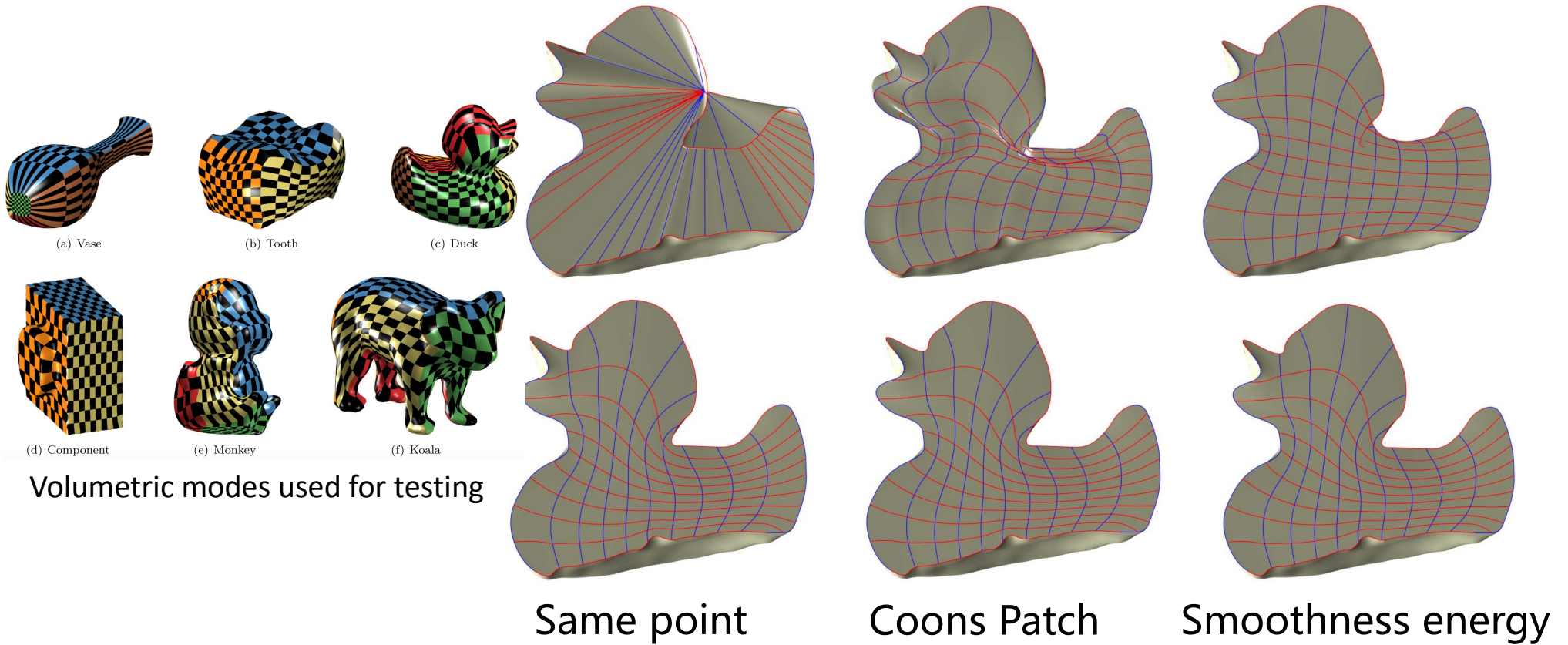


Success rates
PDE (Hinz, J. et al. (2018)): 62.23%
PDE (Ji et al. 2023): 73.80%
Barrier (Ji et al. 2021): **98.36%**
Penalty (Ji et al. 2021): 97.85%

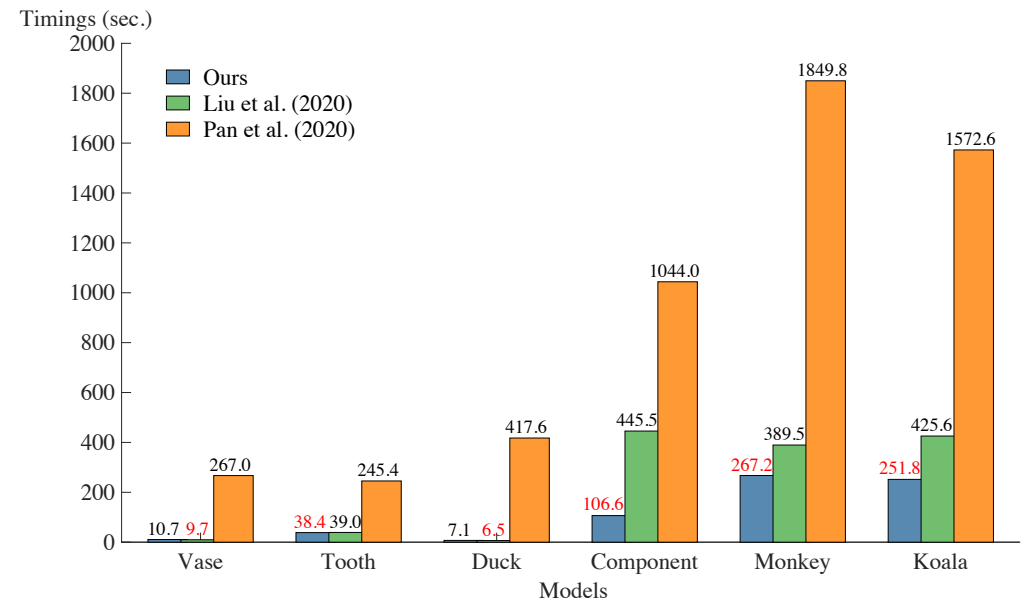
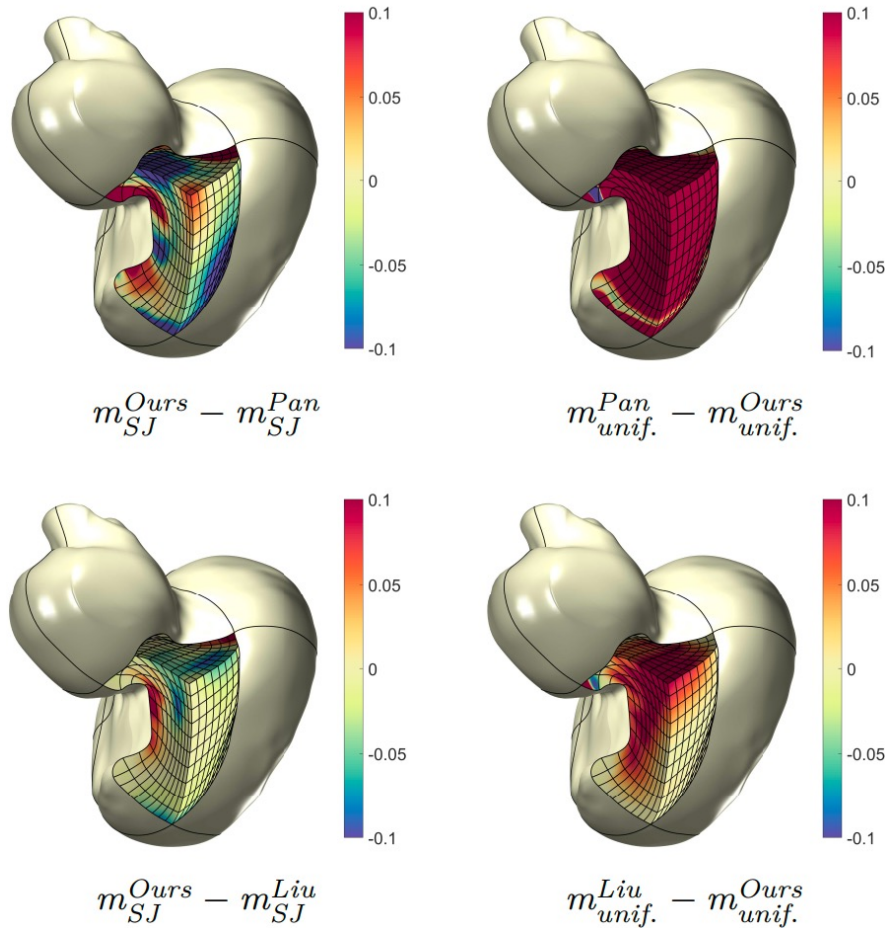
Computational time:
PDE-based: ~0.2 sec
opt-based: ~2 sec

Ji, Y., Möller, M., Verhelst, H.M. (2024). Design Through Analysis. In: Bodnár, T., Galdi, G.P., Nečasová, Š. (eds) Fluids Under Control. Advances in Mathematical Fluid Mechanics. Birkhäuser, Cham.

Strong similarity from different initializations



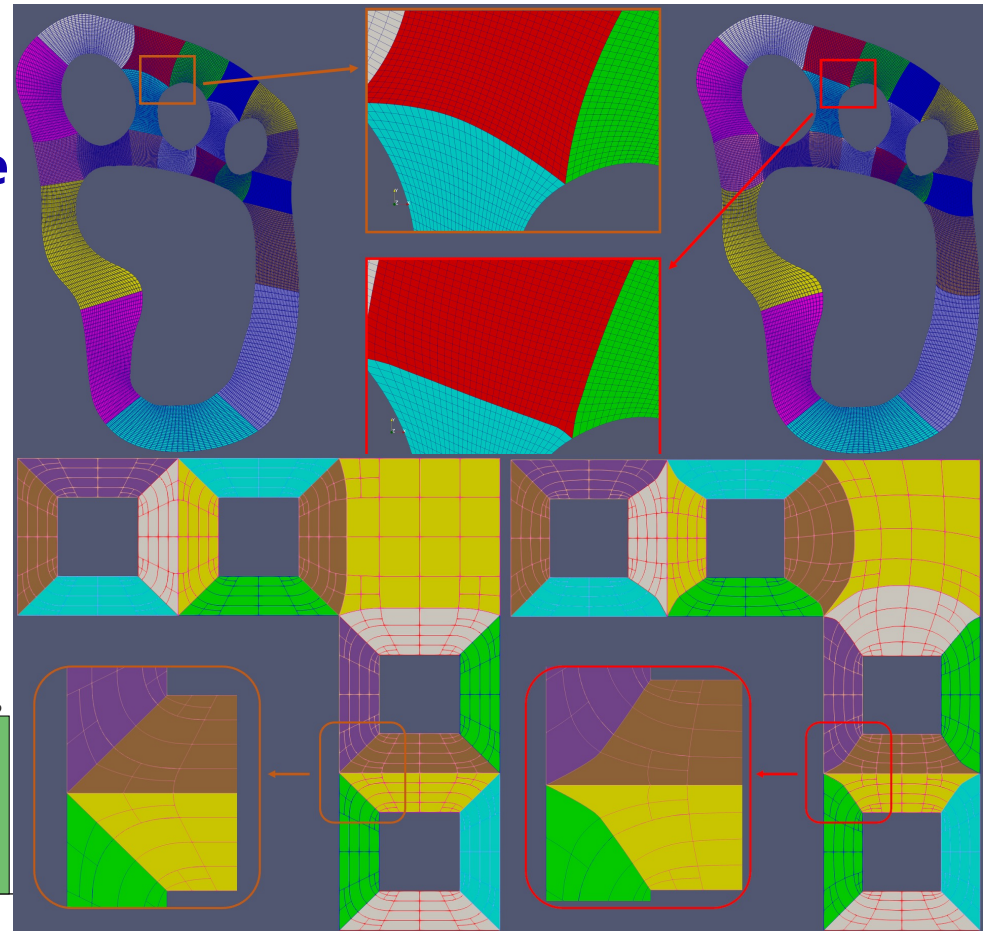
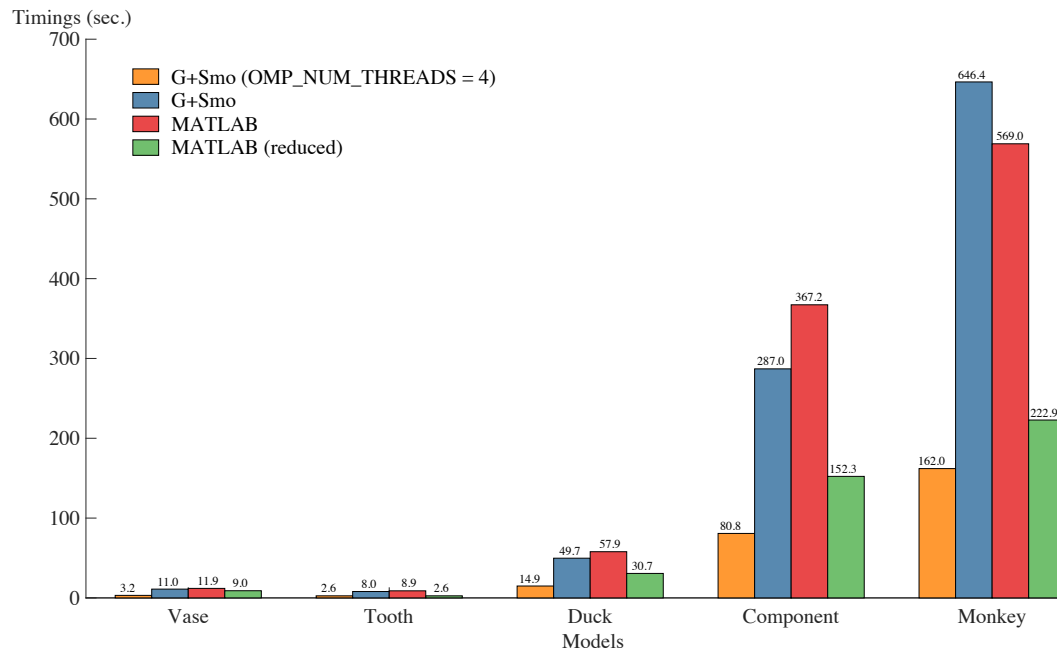
Comparisons with existing methods

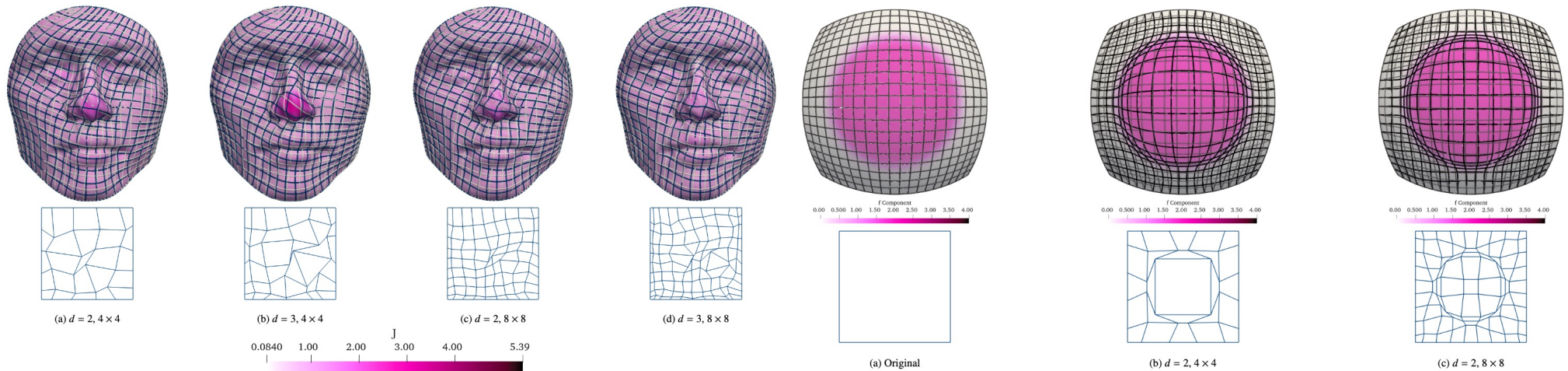


- **Positive values** indicate our method performs better;
- Efficiency comparison (MATLAB vs. C++ (Pan+ & Liu+):
 - Significantly **faster** than Pan et al. (2020);
 - Large-Scale Models: Outperforms Liu et al. (2020).

1. Pan, M., Chen, F., & Tong, W. (2020). CMAME, 359, 112769.
2. Liu, H., Yang, Y., Liu, Y., & Fu, X. M. (2020). CAGD, 79, 101853.

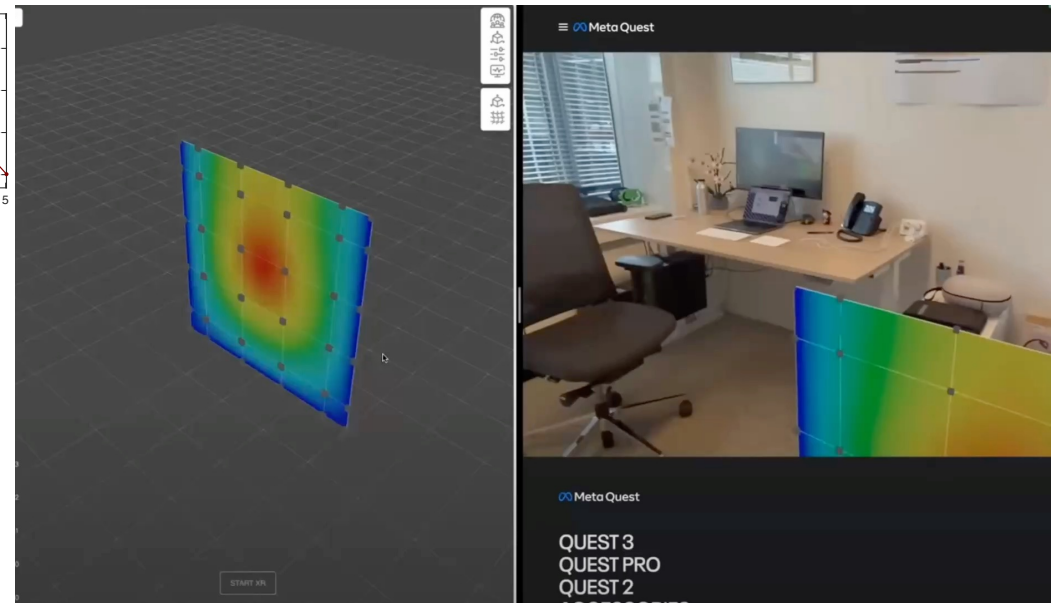
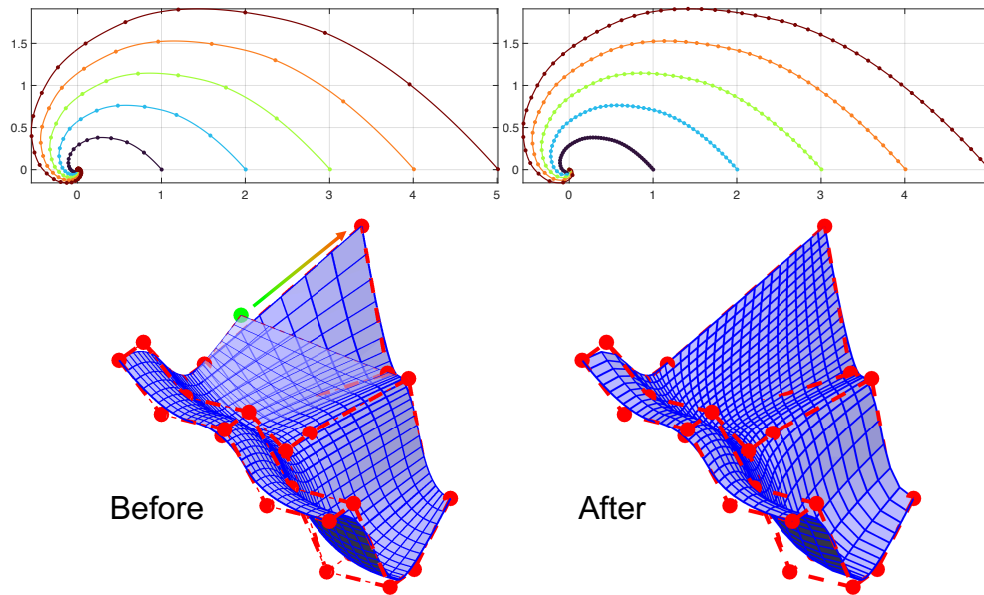
- In our released G+Smo implementation, **3-4x speed-up using OpenMP (!)**;
- Suitable for **multi-patch** and **THB-spline** parameterization;





➤ **A composite spline relocation** framework for surface-based IGA, integrating **harmonic map-based mesh redistribution** with spline-based techniques;

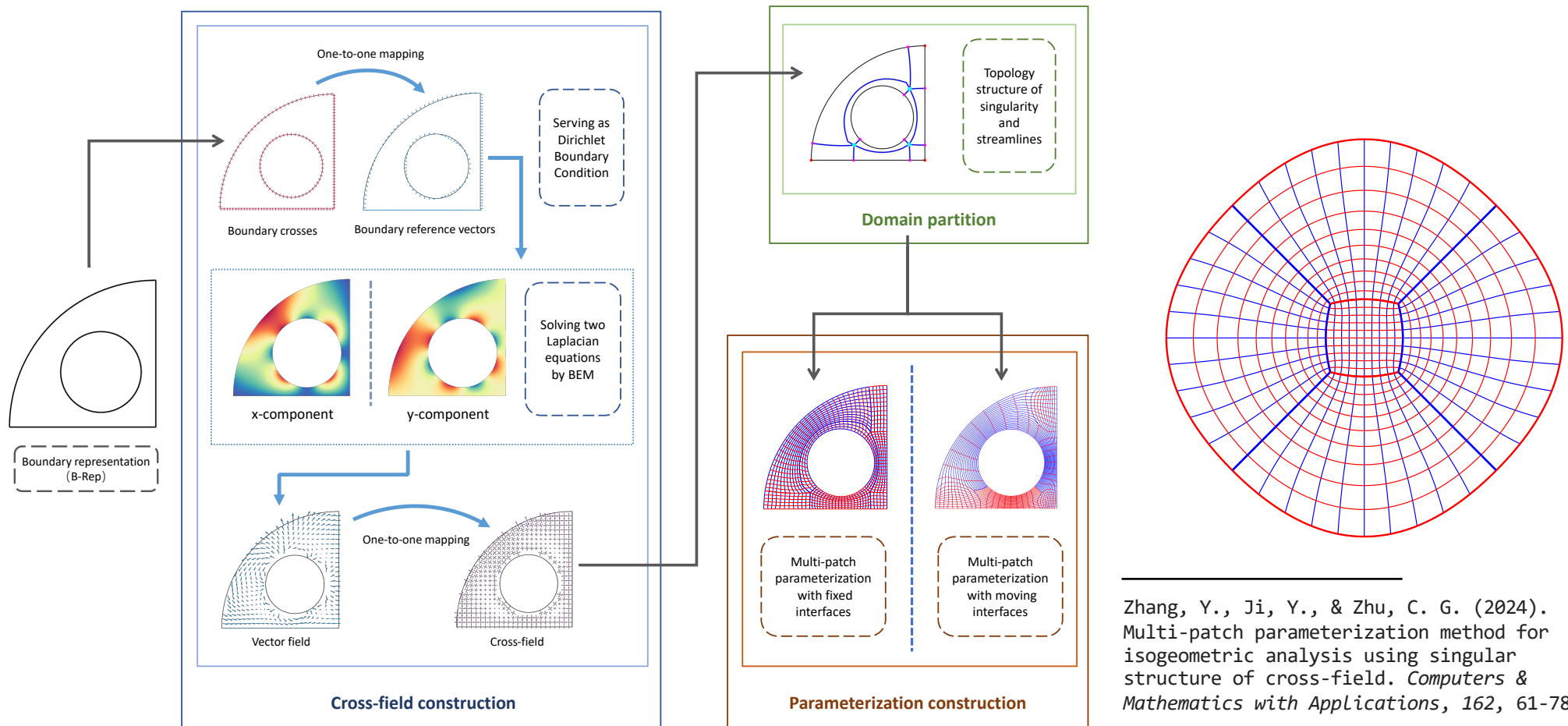
Composite spline relocation for isogeometric analysis on surfaces
joint work with Carlotta Giannelli, Sofia Imperatore, & Hugo Verhelst



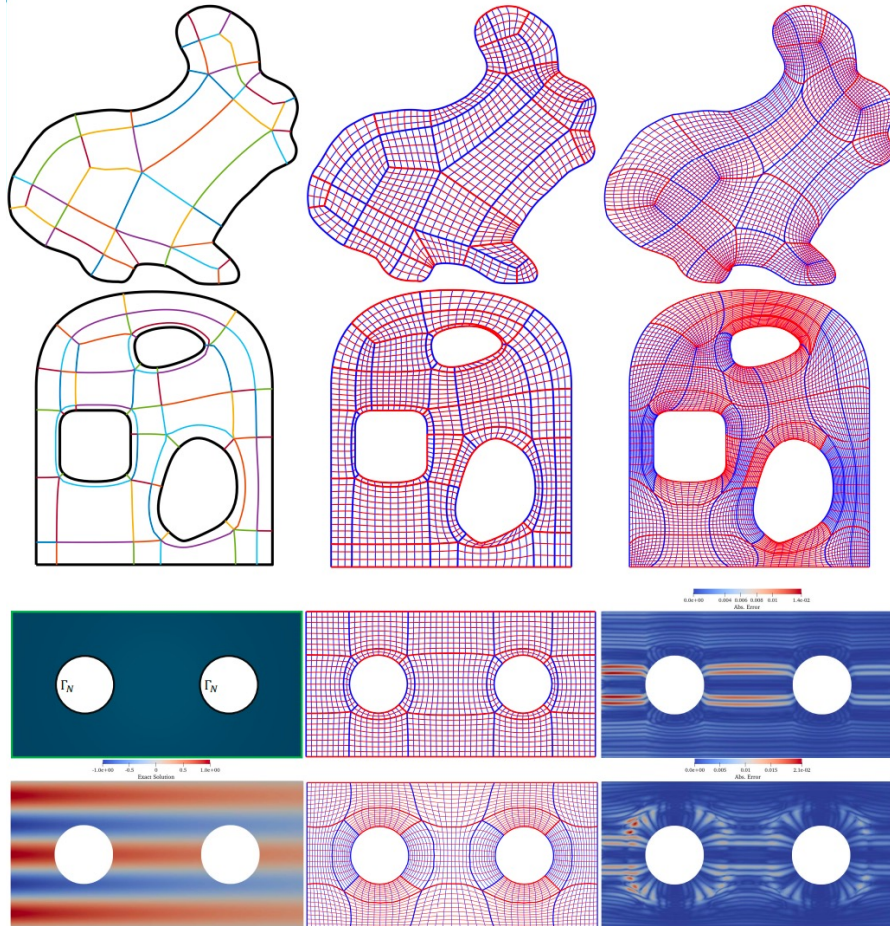
- Curve/surface reparameterization while keeping the geometry;

Design Through Analysis (DTA) Tools - Matthias
Wed. 10:45-12:00 Software session #2

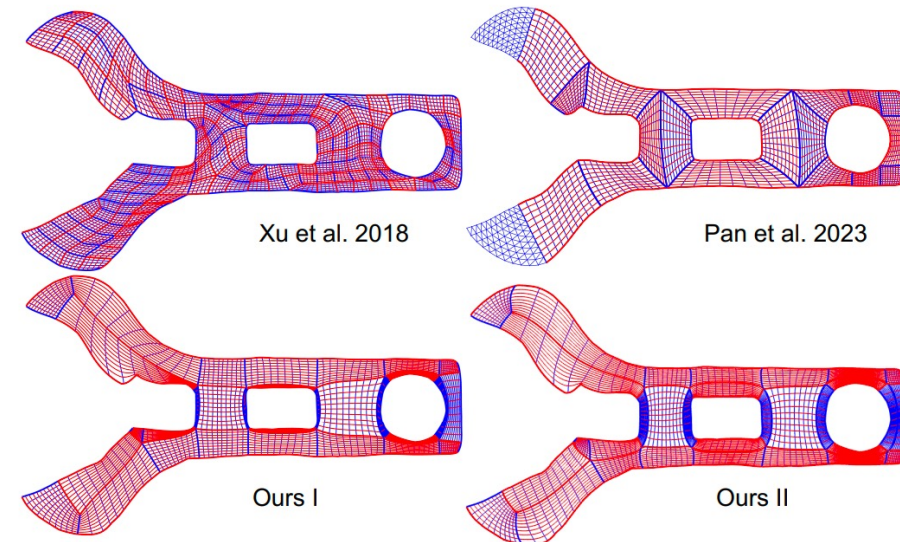
Multi-patch parameterization using cross-field



Zhang, Y., Ji, Y., & Zhu, C. G. (2024). Multi-patch parameterization method for isogeometric analysis using singular structure of cross-field. *Computers & Mathematics with Applications*, 162, 61-78.

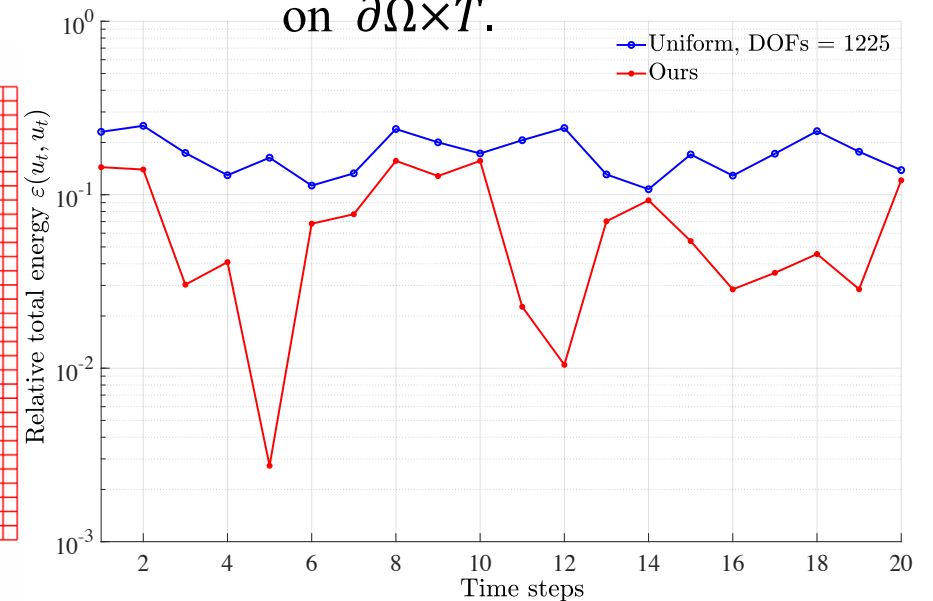
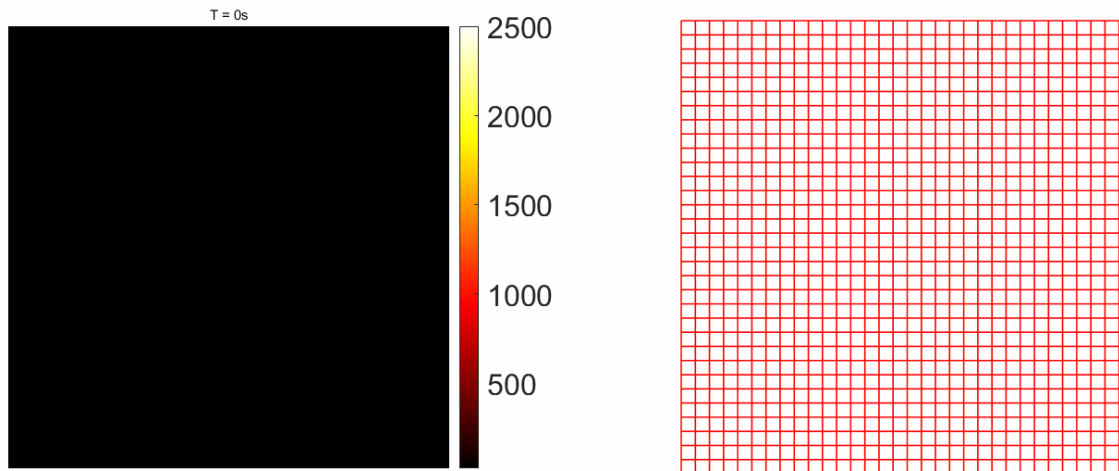


Model	#Patch	Method	$ \mathcal{J} _s$		<i>unif.</i>		
			min.	avg.	min.	avg.	max.
rabbit	33	Coons	-0.8593	0.9628	0.7030	0.9410	1.0524
		fixed-I	0.2204	0.9504	0.6103	0.9544	0.9982
		moving-I	0.02918	0.9283	0.0000	0.9550	1.0000
3 holes	46	Coons	-0.5492	0.9710	0.8008	0.9573	1.0958
		fixed-I	0.1545	0.9716	0.8007	0.9573	0.9978
		moving-I	0.1461	0.9361	0.6791	0.95710	0.9968



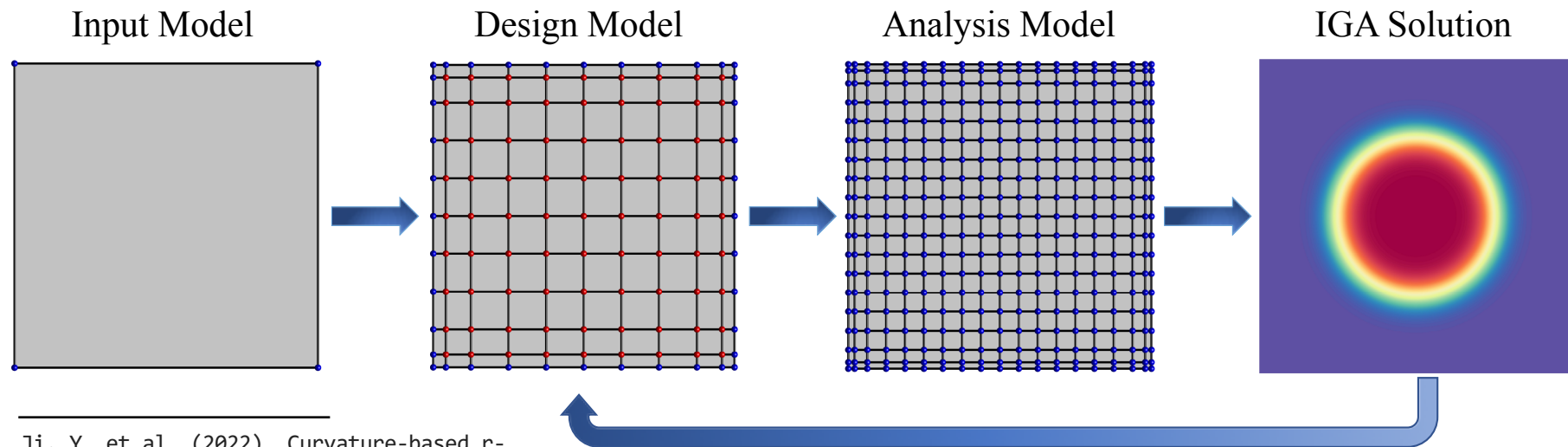
- Consider a two-dimensional linear heat transfer problem with a moving Gaussian heat source:

$$\begin{cases} C_p \rho \frac{\partial u(\mathbf{x}, t)}{\partial t} - \nabla \cdot (\kappa \nabla u(\mathbf{x}, t)) = f(\mathbf{x}, t), & \text{in } \Omega \times T, \\ u(\mathbf{x}, t) = u_0, & \text{in } \Omega, \\ \kappa \nabla u(\mathbf{x}, t) = 0, & \text{on } \partial\Omega \times T. \end{cases}$$



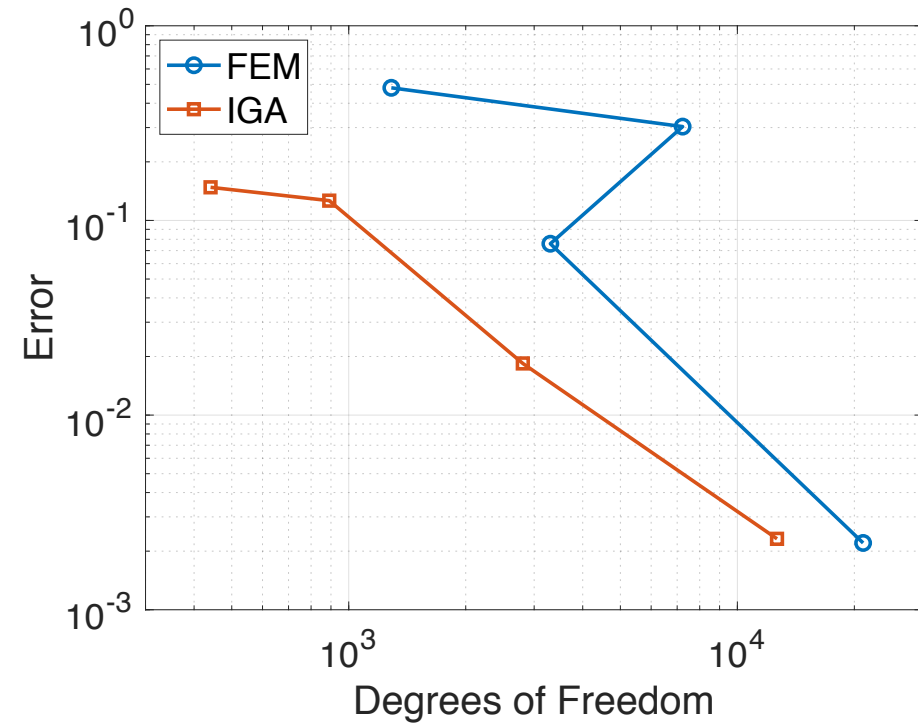
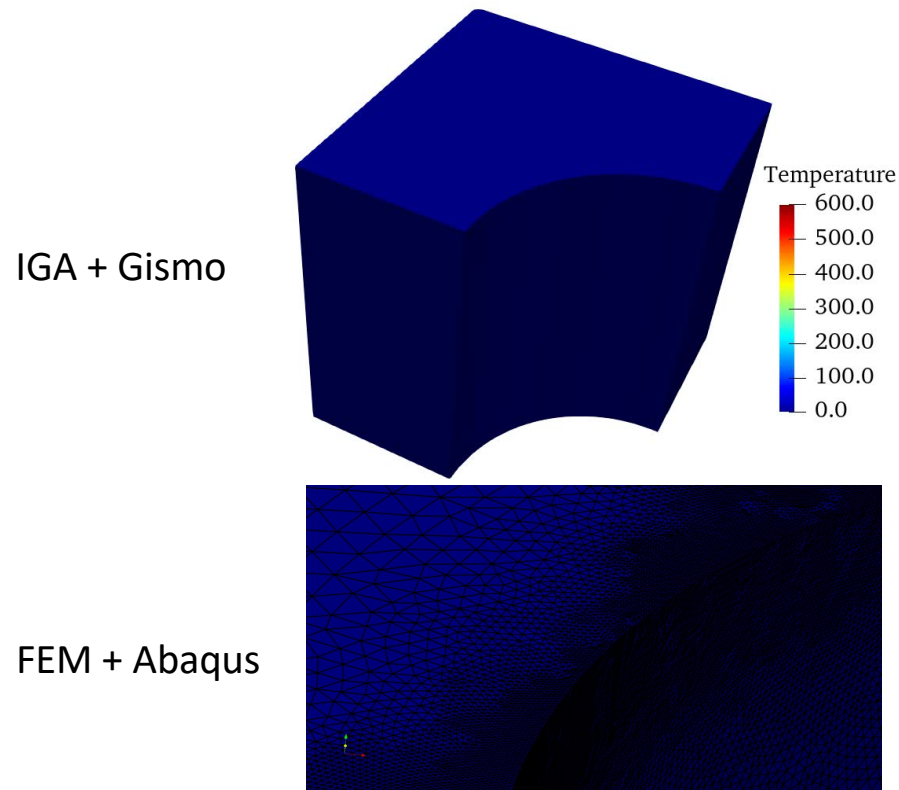
Anisotropic parameterizations are often solution-dependent:

- Need **good numerical solution accuracy** to drive parameterization;
- Adjust as few control points as possible **for high efficiency**;
- **Bi-level strategy**: a coarse level (design model) to update the parameterization for efficiency and a fine level (analysis model) to perform analysis for accuracy.



Ji, Y. et al. (2022). Curvature-based r-adaptive planar NURBS parameterization method for isogeometric analysis using bi-level approach. *Computer-Aided Design*, 150, 103305.

Objective function and sensitivity analysis



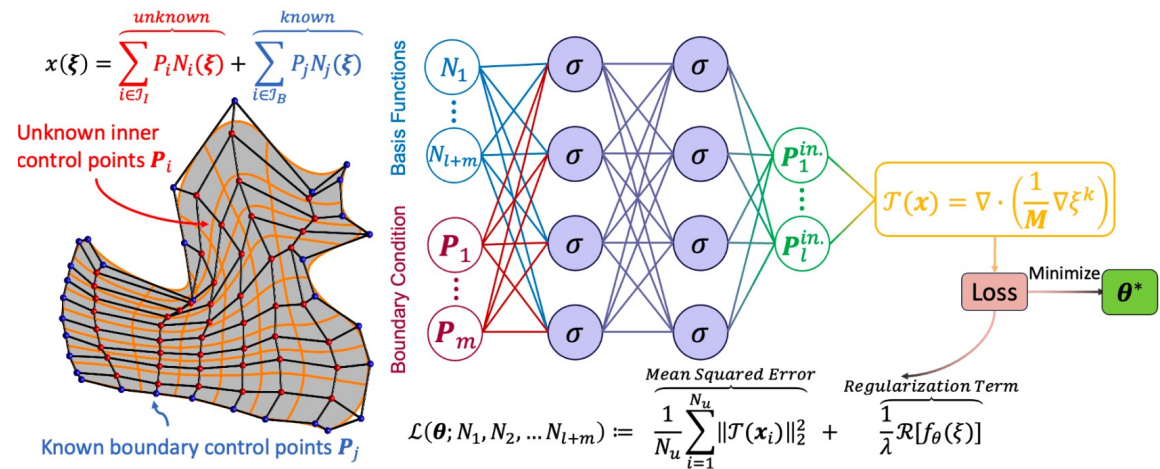
1. Yang, Y., Ji, Y., Möller, M., & Ayas, C. (2025). Computational efficient process simulation of geometrically complex parts in metal additive manufacturing. International Journal of Heat and Mass Transfer, 248, 127059.
2. Semi-analytical heat transfer modeling in metal additive manufacturing using isogeometric analysis (manuscript in preparation)

Network & Training

- 4 fully-connected layers, tanh activations
- Optimizer: Adam, learning rate = 1e-3
- Epochs: 50,000

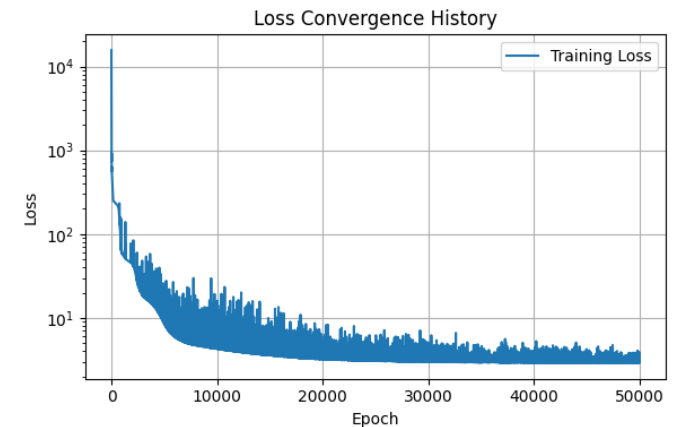
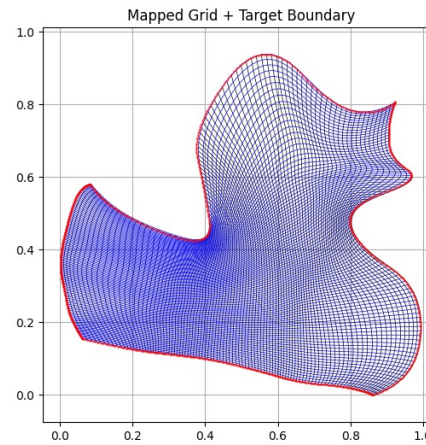
Loss Function

- Interior Loss: Jacobian-regularized Winslow energy $\Rightarrow$ smooth & invertible
- Boundary Loss: MSE between predicted and B-spline-defined boundaries



✓ Powerful auto differentiation tools make it easy to implement;
 ✗ No better than the current "classical" method; **12 min 1.4 sec. (PINN) vs 0.014 sec. (PDE) vs 0.0673 sec. (opt.)**

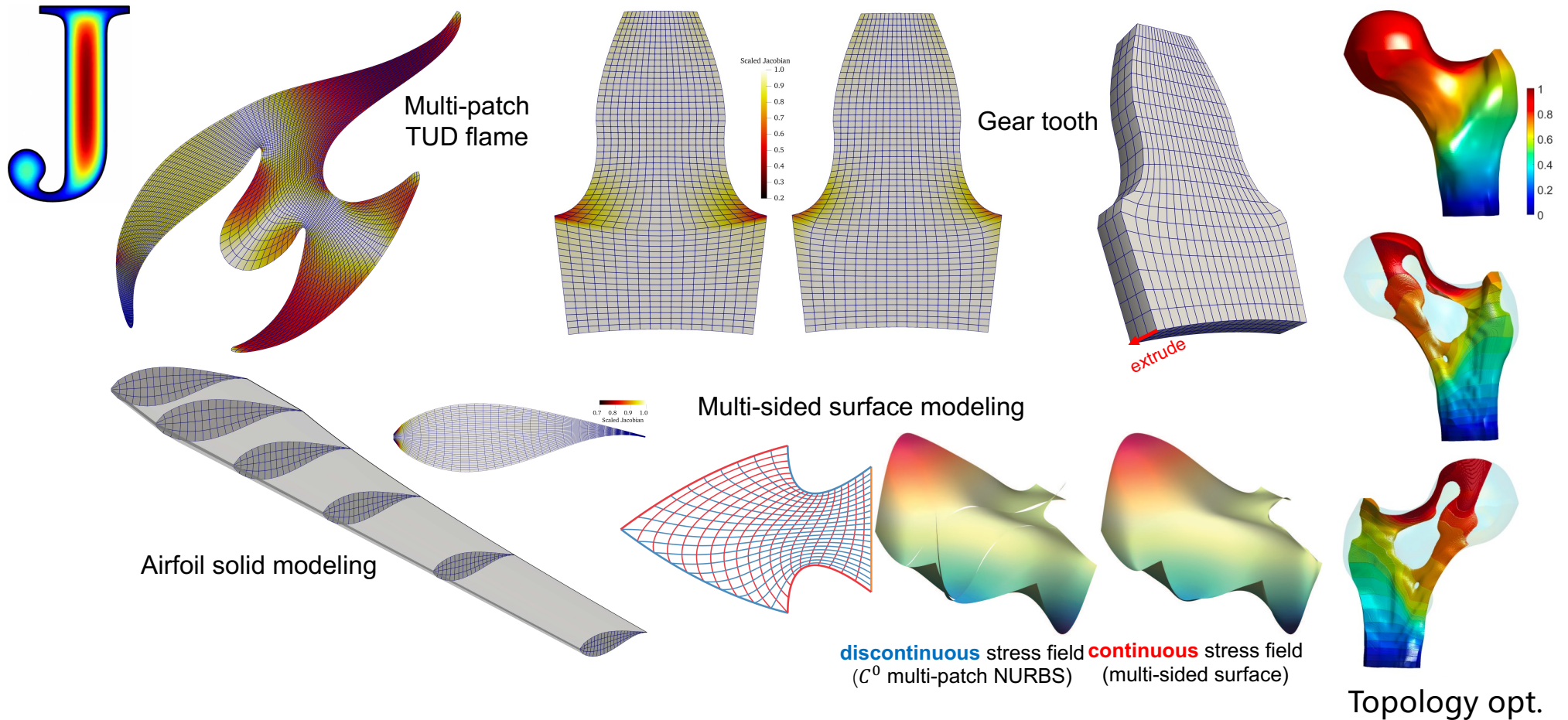
Not an "AI guy" (yet) myself, but very interested in bridging AI with geometry and simulation.

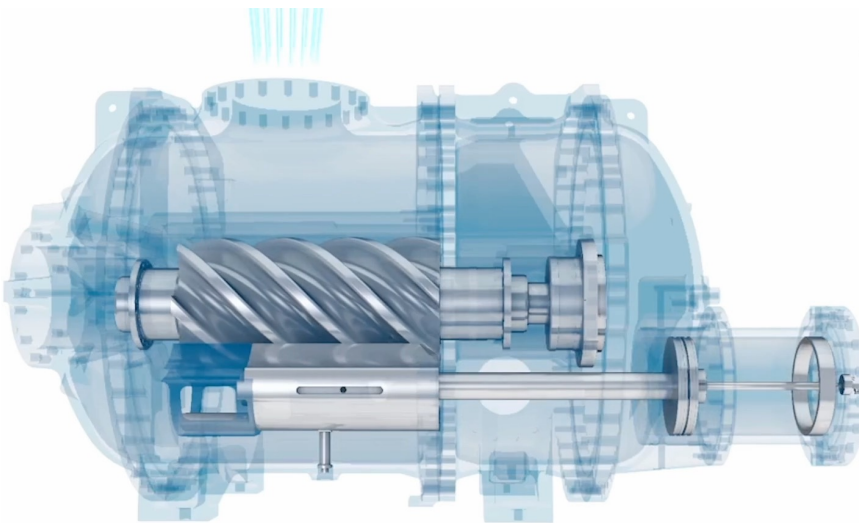




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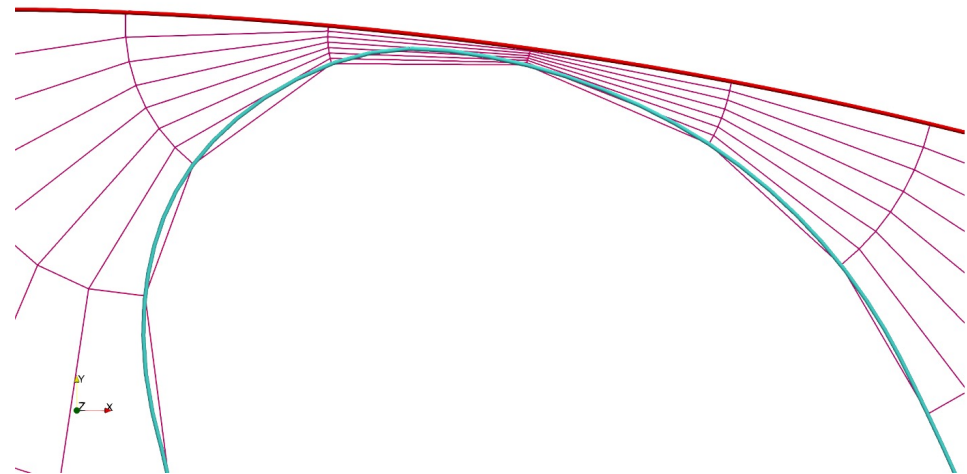




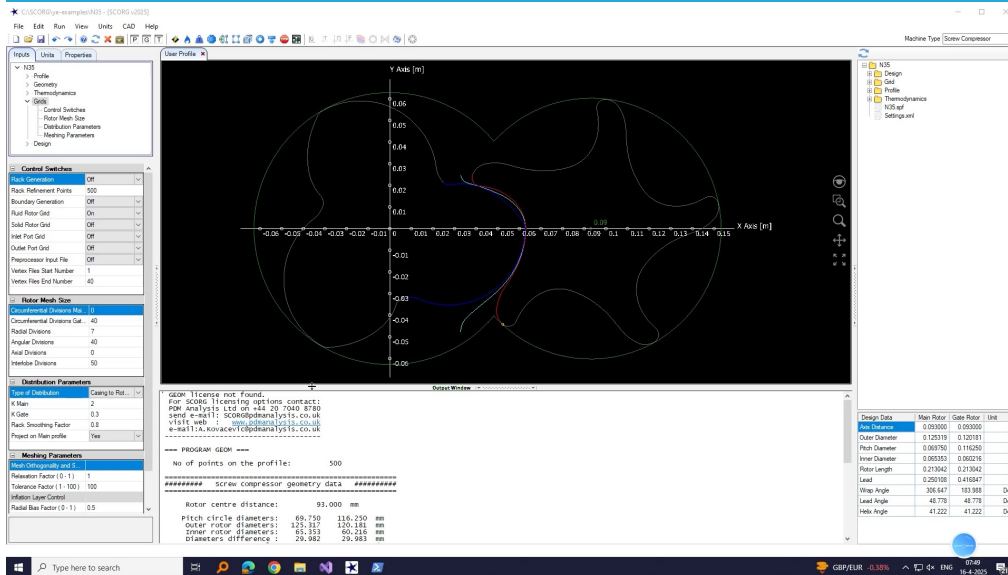
Rotary twin-screw compressor

Source: <https://www.gascompressors.co.uk/technologies/oil-floodedscrew-compressor/>

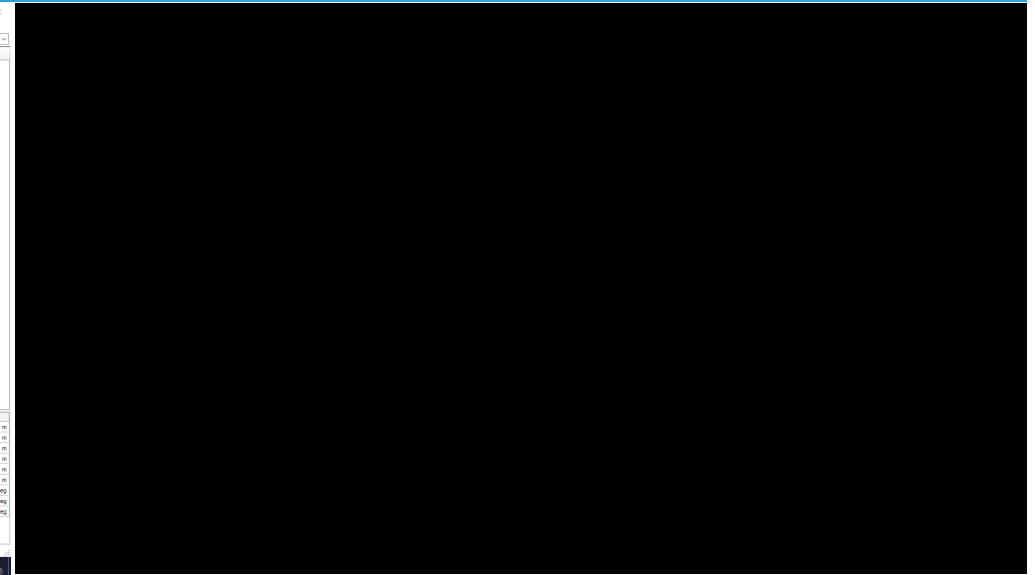
- **Structured mesh generation** is a crucial preprocessing step in the simulation-based analysis of twin-screw machines.
- However, the existing mesh generators typically produce only **linear meshes with straight-sided cells**;
- **Analysis-suitable, high-order NURBS parameterizations.**



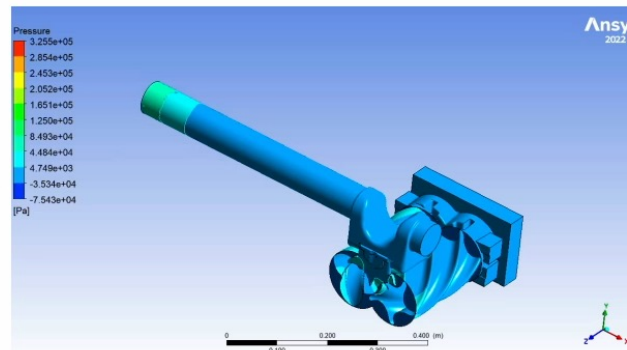
Scorg VS. SplineMesh



Scorg



SplineMesh

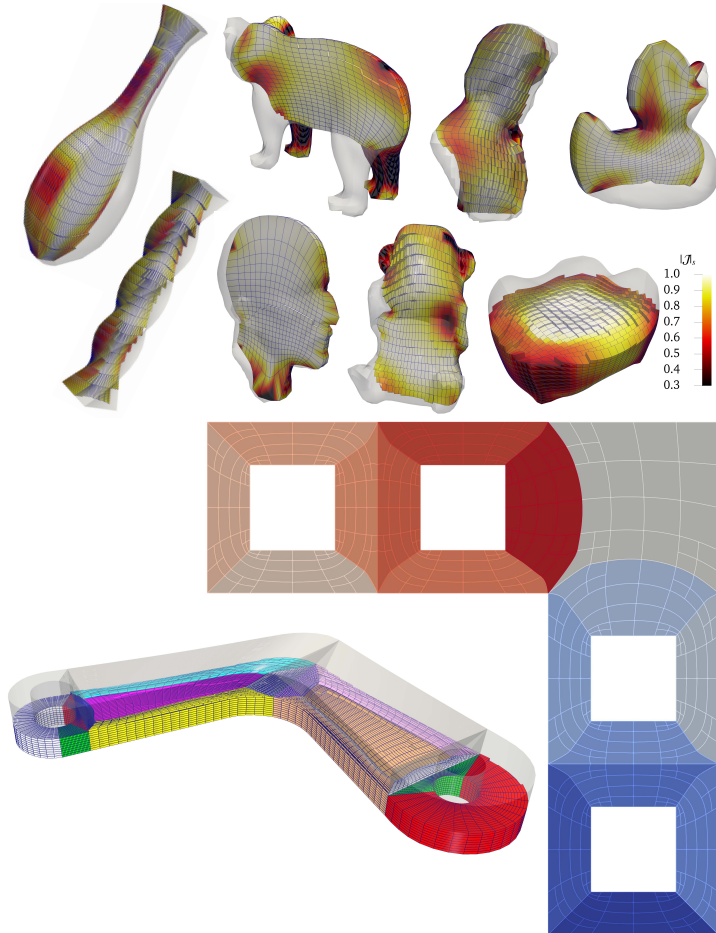


ANSYS CFX simulation

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- **G+Smo offers** three major classes of **parameterization methods**;
- Demonstrates **improved robustness and efficiency** over existing methods;
- Applicability in both **academic** and **real-world industry scenarios**;
- Future Work:
 - Exploring **AI techniques** for geometric modelling and design-through-analysis (DTA) workflows;
 - Developing a user-friendly **graphical user interface** to support broader adoption.



Many thanks for your attention!

Q&A

If you are interested in my research, please feel free to contact me! ;-)

-  Email: y.ji-1@tudelft.nl
-  GitHub: [jiyess](https://github.com/jiyess)
-  Homepage: <https://jiyess.github.io/>