



IGA-suitable parameterization: Advancements inside and outside G+Smo

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Agenda

- 1 Domain parameterization problem
- 2 Inside G+Smo
- 3 Applications
- 4 Outside G+Smo
- 5 Conclusions

Agenda

① Domain parameterization problem

② Inside G+Smo

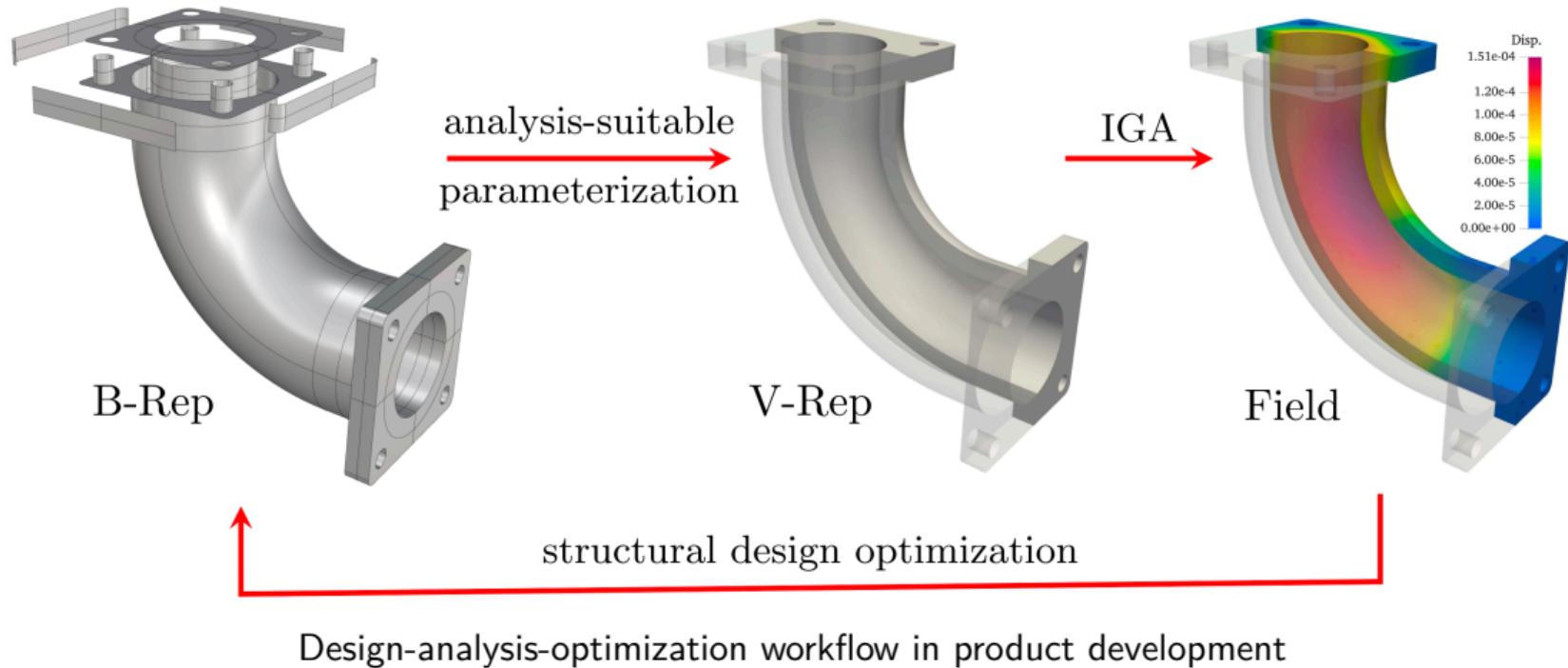
③ Applications

④ Outside G+Smo

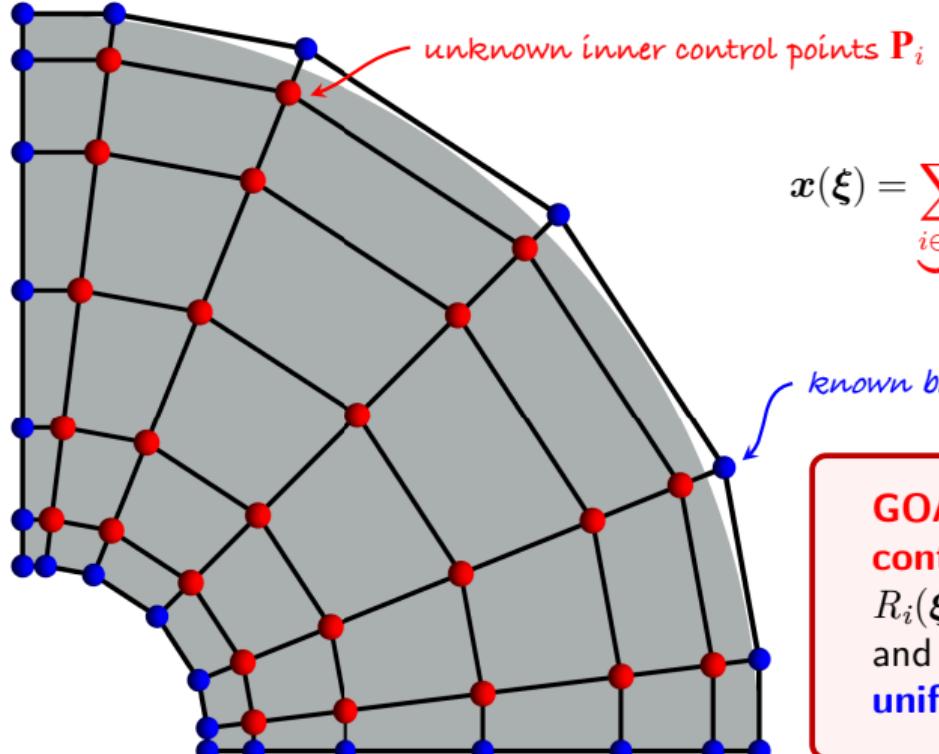
⑤ Conclusions

Domain parameterization for isogeometric analysis

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



Problem statement: domain parameterization



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

known boundary control points \mathbf{P}_j

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

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

Agenda

① Domain parameterization problem

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⑤ Conclusions

Inside G+Smo: as_parameterization_example.cpp

Class Name	Brief Description
gsBarrierCore<d,T>	Core class for AS-parameterizations using various approaches.
gsBarrierPatch<d,T>	Data and geometry preprocessing for gsBarrierCore<d,T>.
gsHLBFGS<T>	Wrapper for the Hybrid Low-storage BFGS optimization solver.
AndersonAcceleration<T>	Anderson acceleration solver and its preconditioned variants.
preAAParam<T>	Parameters for the preconditioned AA solver.

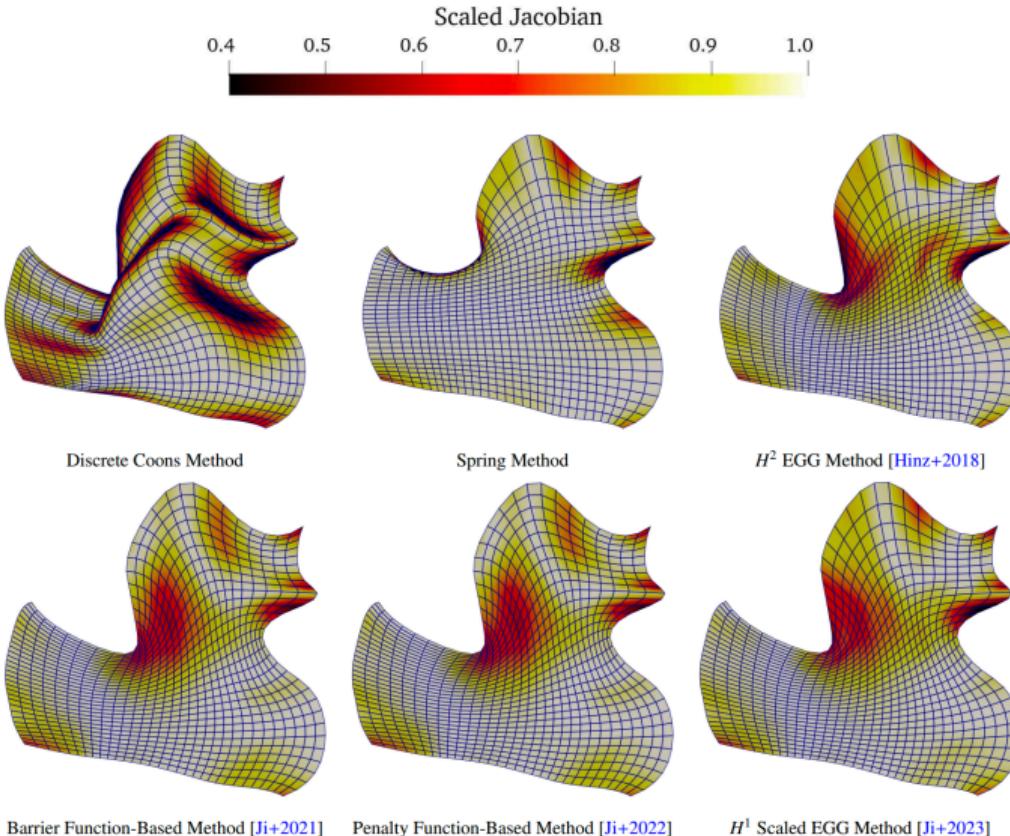
- Example file: `as_parameterization_example.cpp`
- Test inputs: `filedata/breps`
 - 2D case: `filedata/breps/2D/duck_BRep.xml`
 - 3D case: `filedata/breps/3D/duck_BRep.xml`
 - multi-patch case: `filedata/breps/other/TUDflame.xml`, credits to Hugo Verhelst.

Inside G+Smo: as_parameterization_example.cpp

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- Example file: `as_parameterization_example.cpp`
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 - 3D case: `filedata/breps/3D/duck_BRep.xml`
 - multi-patch case: `filedata/breps/other/TUDflame.xml`, credits to Hugo Verhelst.
- **Many available parameterization methods in G+Smo**
 - Not get too much into algorithm details today. ;-)

Available methods in G+Smo



- 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]

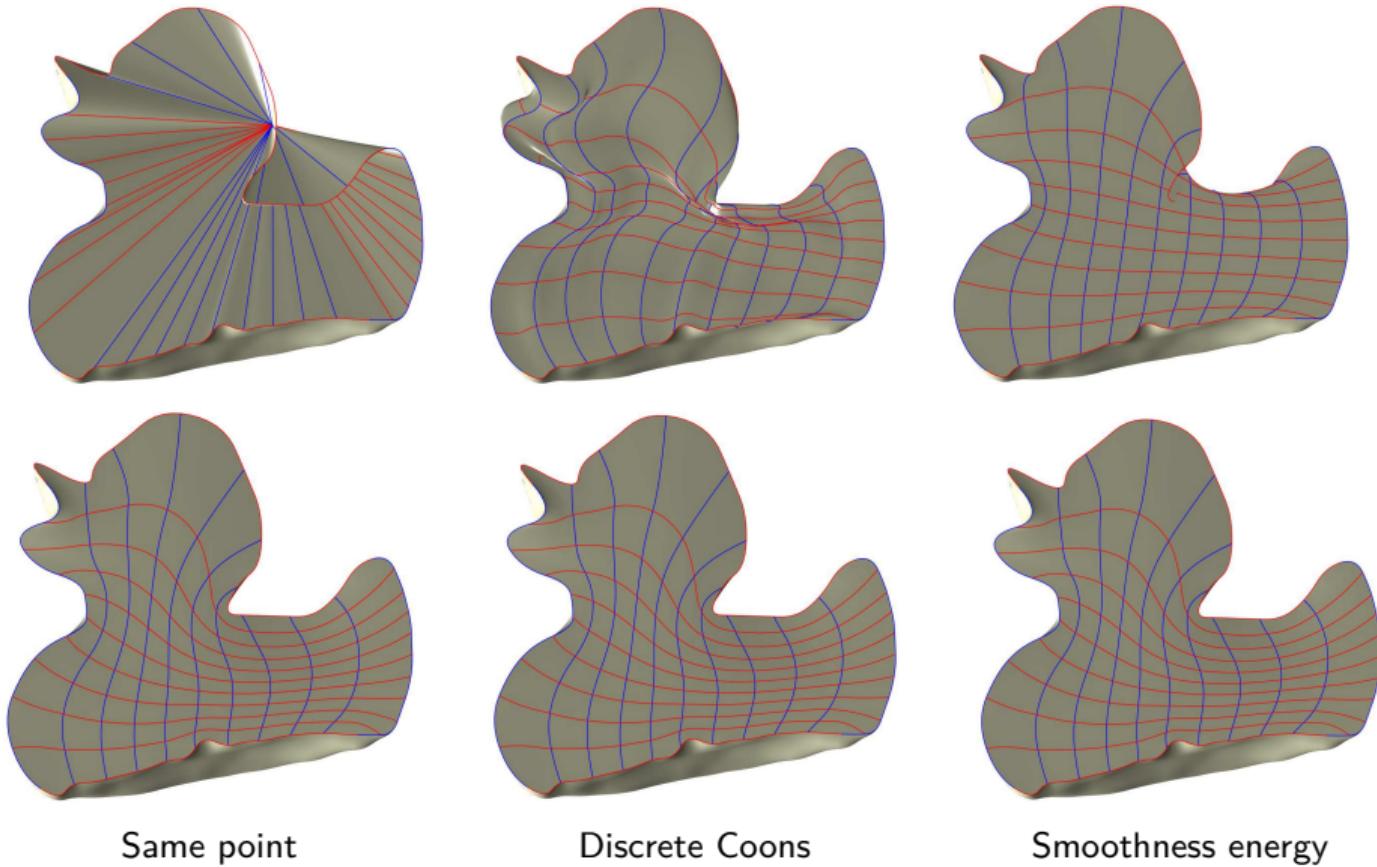
[1] Ji, Y. et al. (2021). Constructing high-quality ...

[2] Ji, Y. et al. (2022). Penalty function-based ...

[3] Hinz, J. et al. (2018). Elliptic grid generation ...

[4] Ji, Y. et al. (2023). On an improved PDE-based ...

Strong similarity from different initializations

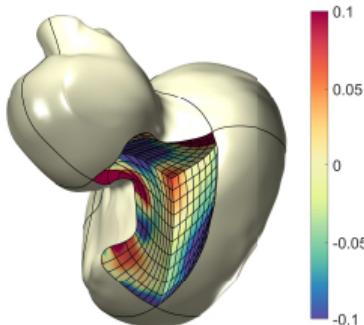


Same point

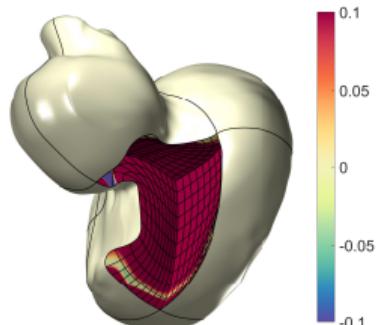
Discrete Coons

Smoothness energy

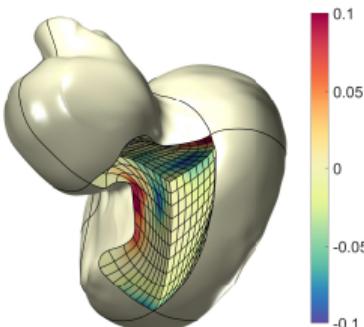
Parameterization Quality Comparison



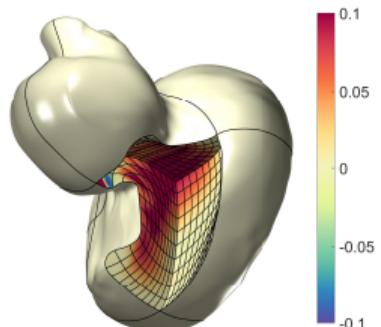
$$m_{SJ}^{Ours} - m_{SJ}^{Pan}$$



$$m_{unif.}^{Pan} - m_{unif.}^{Ours}$$



$$m_{SJ}^{Ours} - m_{SJ}^{Liu}$$



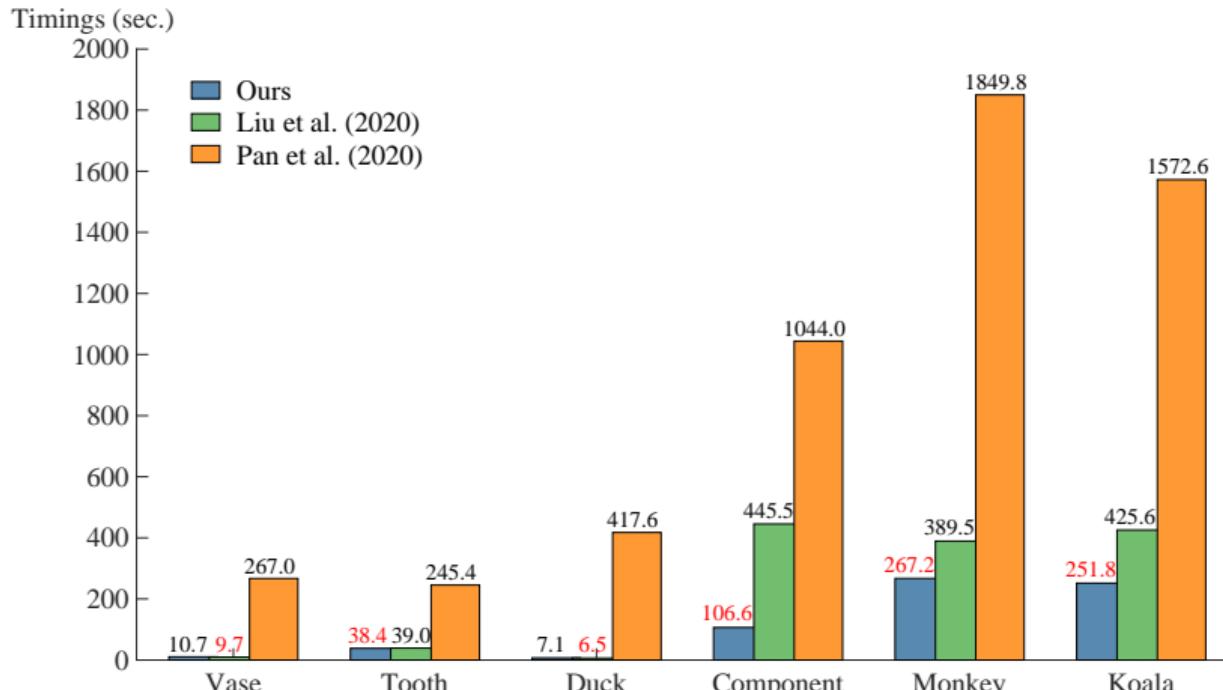
$$m_{unif.}^{Liu} - m_{unif.}^{Ours}$$

- Comparisons to Pan et al. 2020 [1] and Liu et al. 2020 [2]
- Positive values (red regions) indicate our method performs better

- [1] Pan, M., Chen, F., & Tong, W. (2020). Volumetric spline parameterization for isogeometric analysis. *Computer Methods in Applied Mechanics and Engineering*, 359, 112769.
- [2] Liu, H., Yang, Y., Liu, Y., & Fu, X. M. (2020). Simultaneous interior and boundary optimization of volumetric domain parameterizations for IGA. *Computer Aided Geometric Design*, 79, 101853.

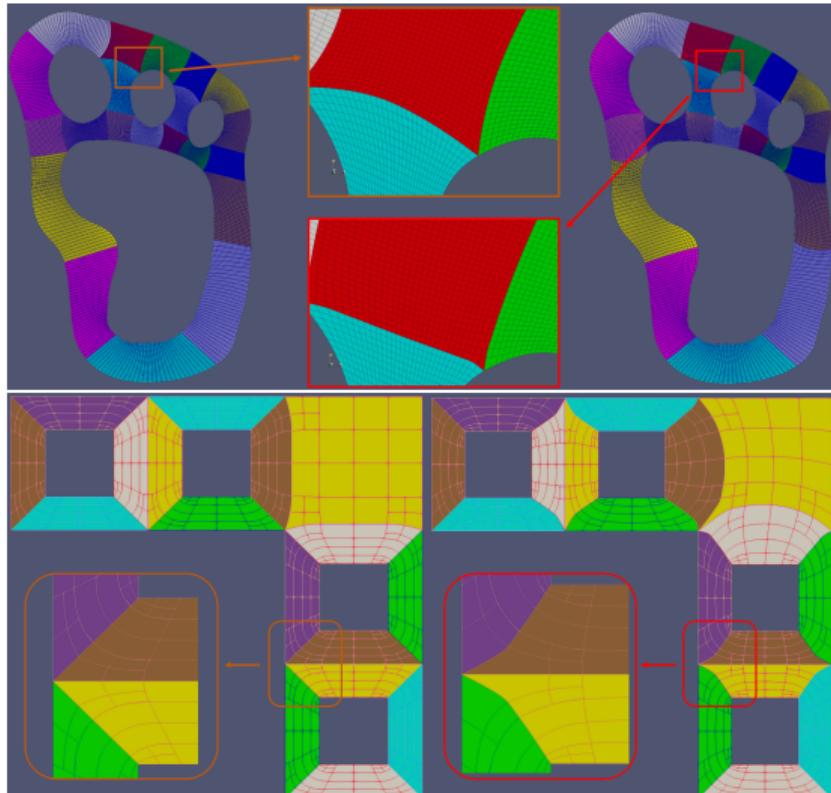
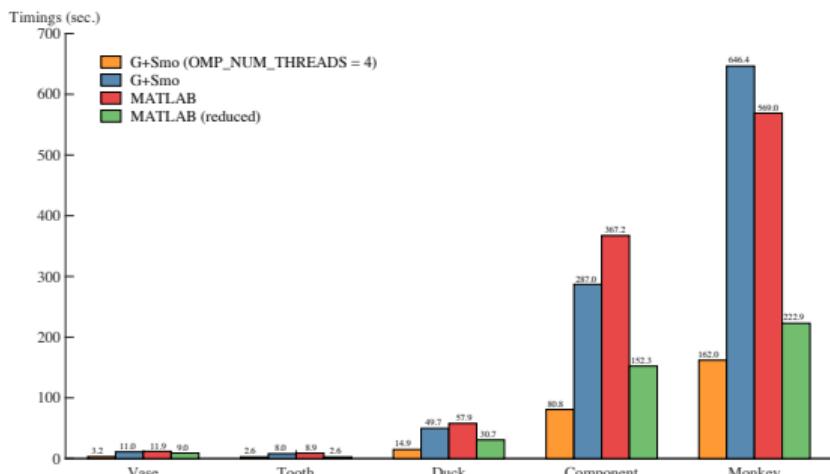
Efficiency Comparison

- Significantly **faster** than Pan et al. (2020);
- **Large-Scale Models:** Outperforms Liu et al. (2020);
- **Implementation:** MATLAB (ours) vs. C++ (Pan et al. & Liu et al.).



G+Smo Implementation

- In our released G+Smo implementation, **3-4x speed-up**;
- Suitable for **multi-patch** and **THB-spline** parameterization;
- **Precomputation and fast matrix assembly in progress.**



Elliptic Grid Generation (EGG) & Its H^2 Discretization

- To compute a harmonic mapping $\mathbf{x} : \hat{\Omega} \rightarrow \Omega$ by solving Laplace equations:

$$\begin{cases} \nabla \cdot \nabla \xi(x, y) = 0 \\ \nabla \cdot \nabla \eta(x, y) = 0 \end{cases} \quad \text{s.t. } \mathbf{x}^{-1}|_{\partial\Omega} = \partial\hat{\Omega}.$$

- **Existence and uniqueness**, and unique solution \mathbf{x}^{-1} offers a **one-to-one mapping** (Radó-Kneser-Choquet theorem).

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- Nonlinear vector-valued second-order PDE [1]

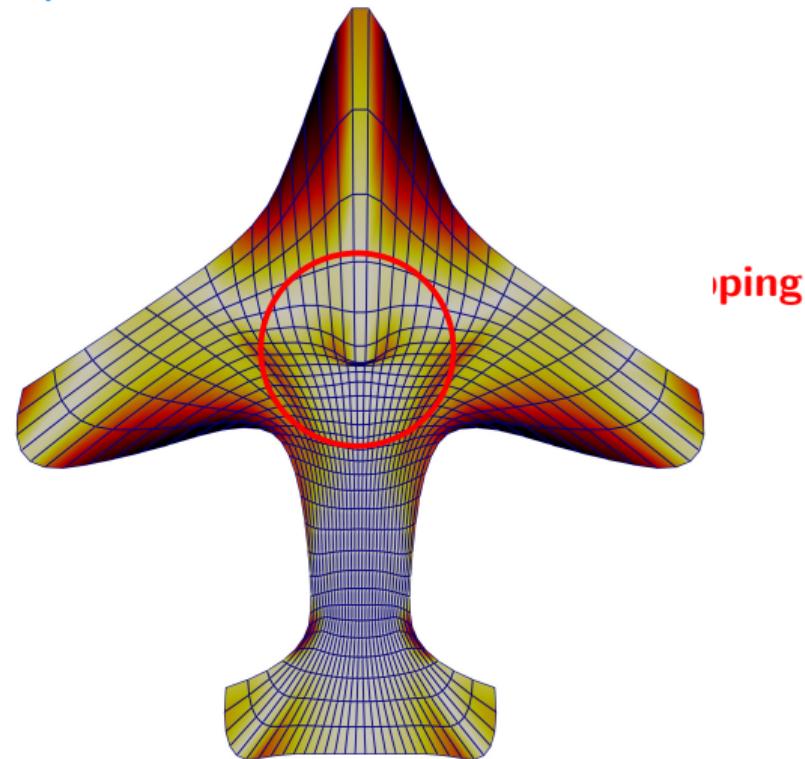
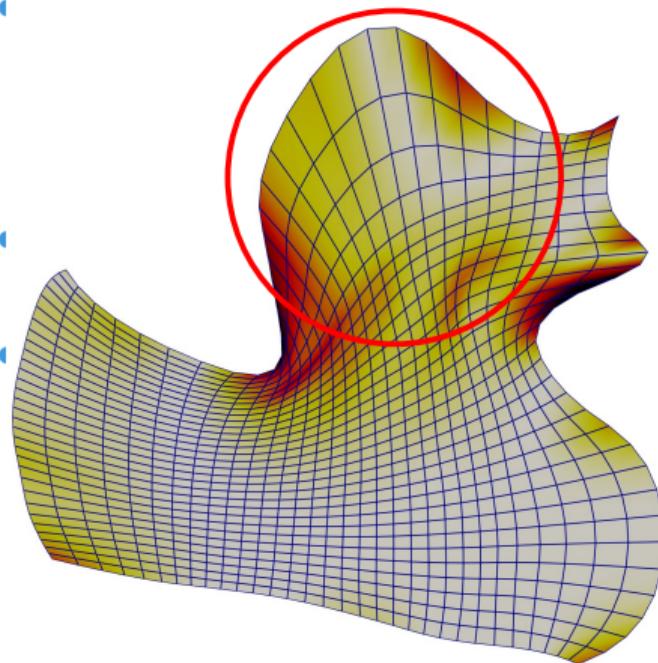
$$\forall R_i \in \Sigma_0 : \begin{cases} \int_{\hat{\Omega}} \mathbf{R} \tilde{\mathcal{L}}x \, d\hat{\Omega} = \mathbf{0}, \\ \int_{\hat{\Omega}} \mathbf{R} \tilde{\mathcal{L}}y \, d\hat{\Omega} = \mathbf{0}, \end{cases} \quad \text{s.t. } \mathbf{x}|_{\partial\hat{\Omega}} = \partial\Omega,$$

where

$$\tilde{\mathcal{L}} = \frac{\mathcal{L}}{g_{11} + g_{22}}, \quad \text{and } \mathcal{L} = g_{22} \frac{\partial^2}{\partial \xi^2} - 2g_{12} \frac{\partial^2}{\partial \xi \partial \eta} + g_{11} \frac{\partial^2}{\partial \eta^2}.$$

[1] Hinz, J. P. (2020). PDE-Based Parameterization Techniques for Isogeometric Analysis Applications. TU Delft doctoral dissertation.

Elliptic Grid Generation (EGG) & Its H^2 Discretization



- **Non-uniform** elements appear, may even be **non-bijective**.

[1] Hinz, J. P. (2020). PDE-Based Parameterization Techniques for Isogeometric Analysis Applications. TU Delft doctoral dissertation.

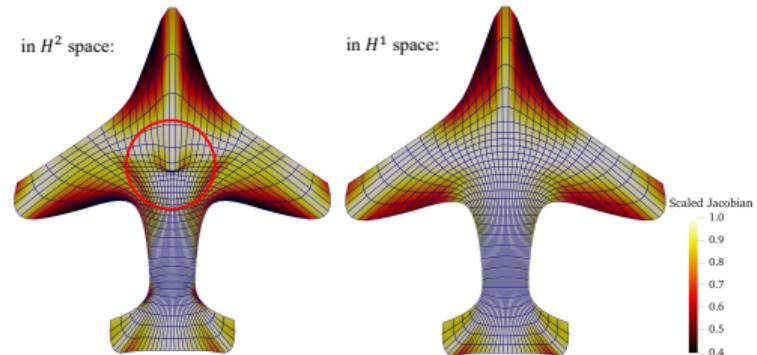
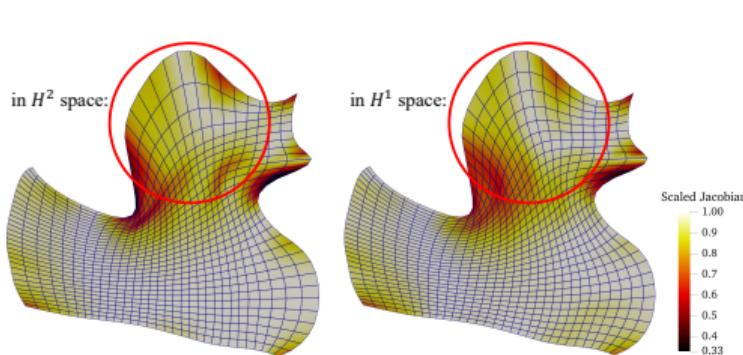
Quasi-Hamonic Mapping

- We introduce a monitor function \mathbb{A} into our governing equation:

$$-\nabla \cdot (\mathbb{A} \nabla \xi) = 0.$$

where the monitor function is defined as

$$\mathbb{A} = \begin{bmatrix} \frac{1}{|\mathcal{J}|} & 0 \\ 0 & \frac{1}{|\mathcal{J}|} \end{bmatrix}.$$



Discretization in Sobolev space H^1

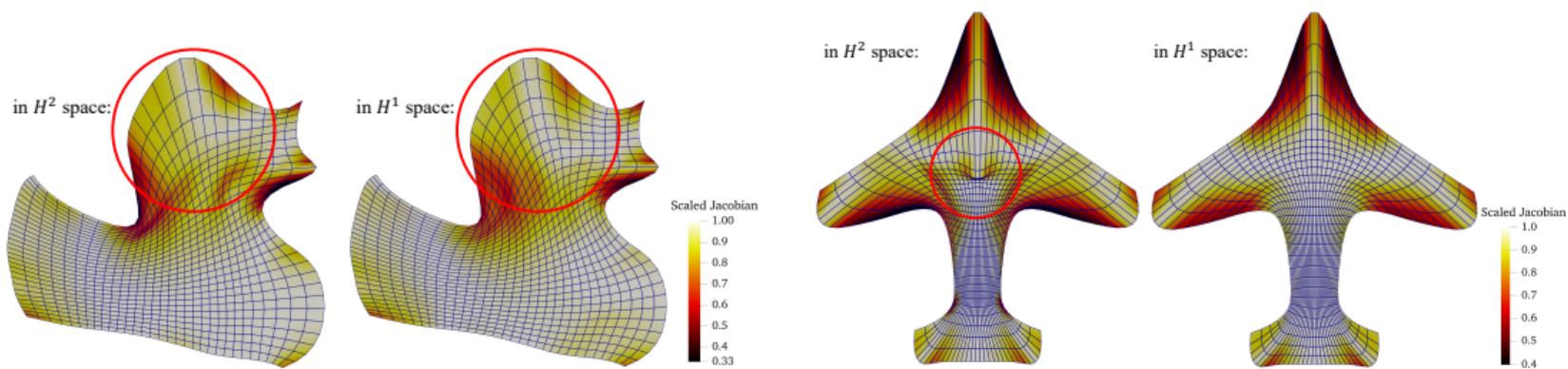
- Variational formulation in Sobolev space H^1 reads

$$\forall R_i \in \Sigma_0 : \begin{cases} \mathbf{F}_{H^1}^x = \mathbf{0}, \\ \mathbf{F}_{H^1}^y = \mathbf{0}, \end{cases} \text{ s.t. } \mathbf{x}|_{\partial\hat{\Omega}} = \partial\Omega.$$

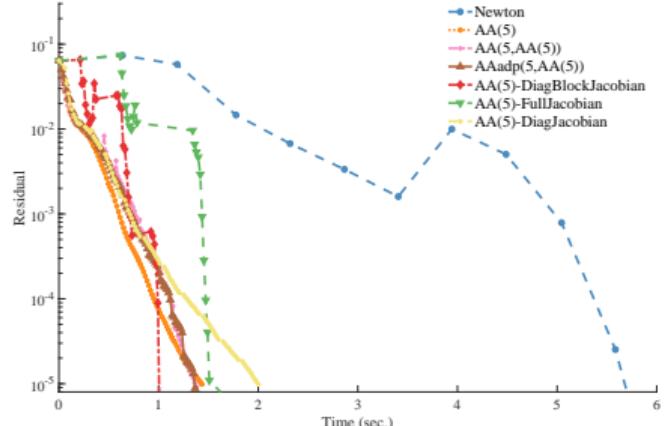
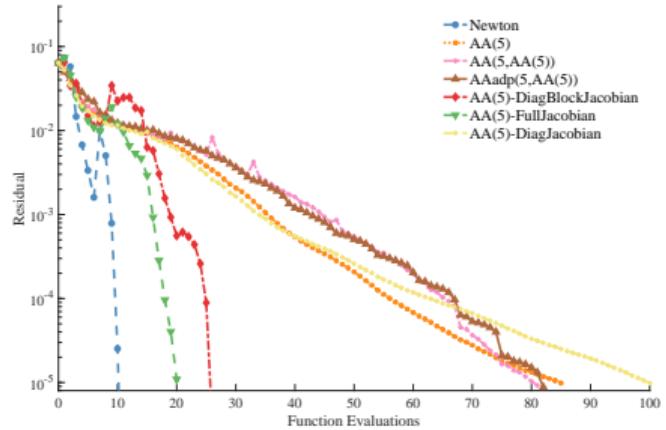
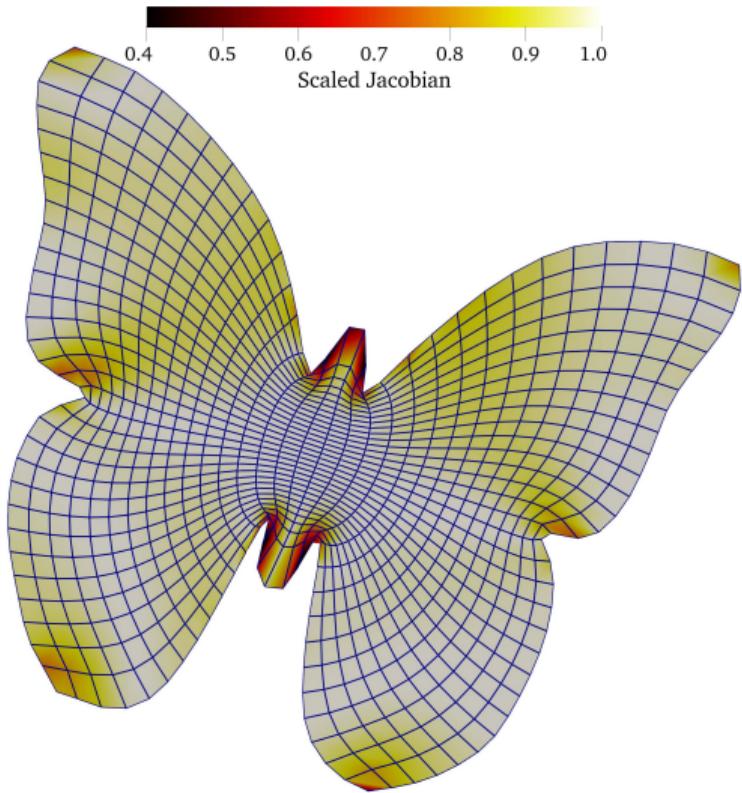
where

$$\mathbf{F}_{H^1}^x = \int_{\hat{\Omega}} \nabla_x \mathbf{R} \cdot \nabla_x \xi \, d\hat{\Omega}, \quad \mathbf{F}_{H^1}^y = \int_{\hat{\Omega}} \nabla_x \mathbf{R} \cdot \nabla_x \eta \, d\hat{\Omega}.$$

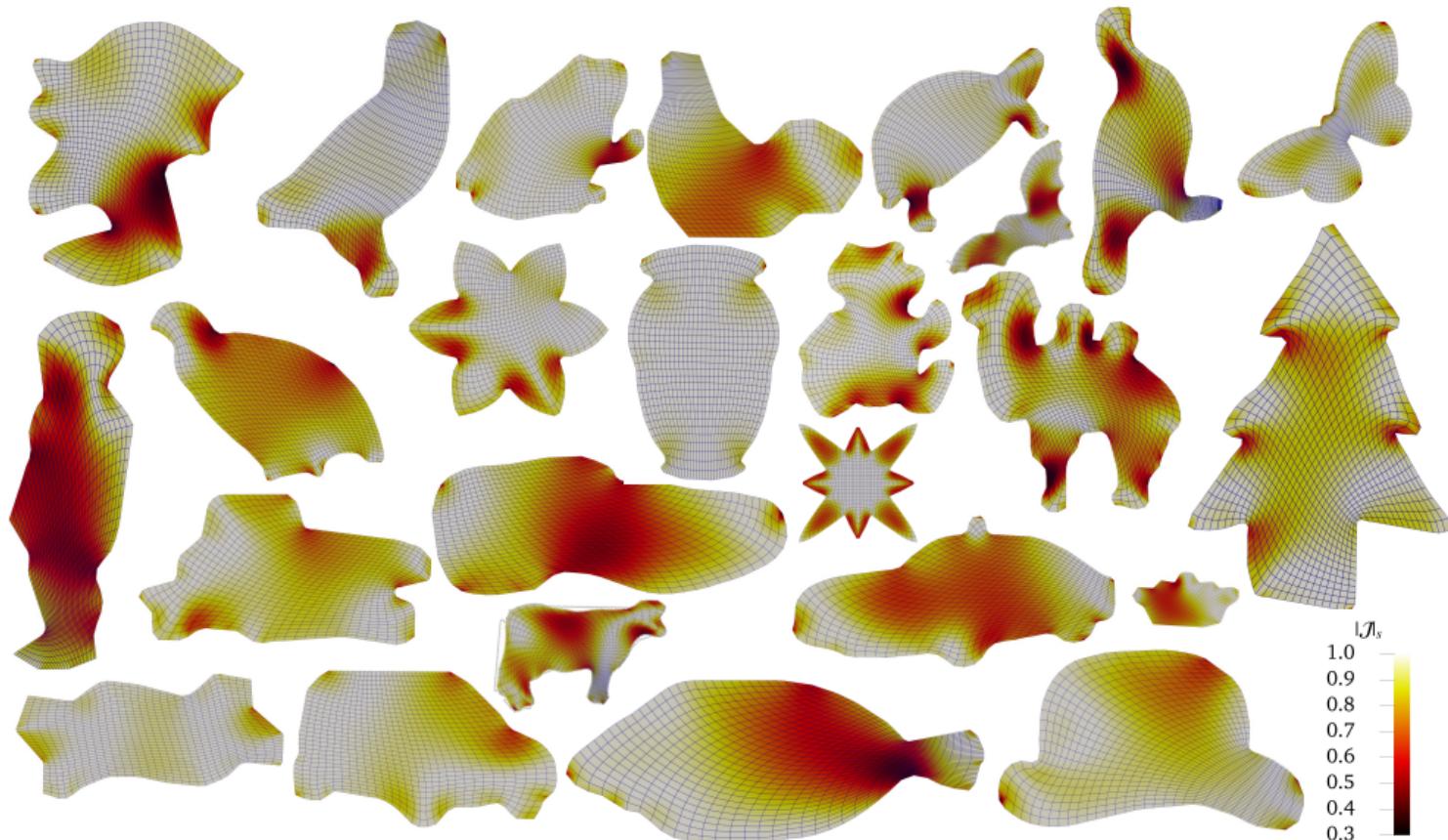
- Needs to solve a **nonlinear system**.



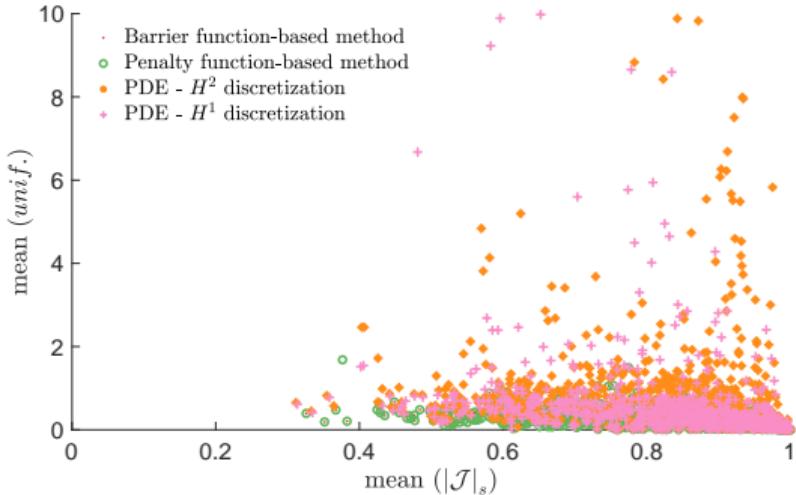
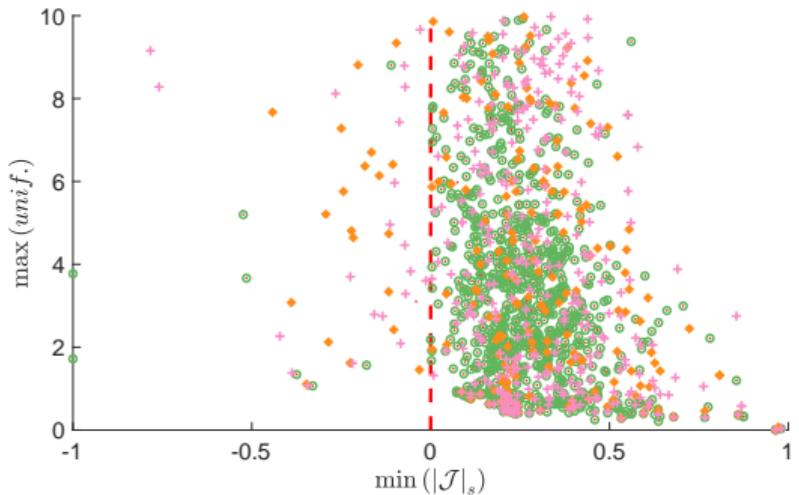
Preconditioned Anderson acceleration solver



Planar Parameterization Test Dataset (977 models)



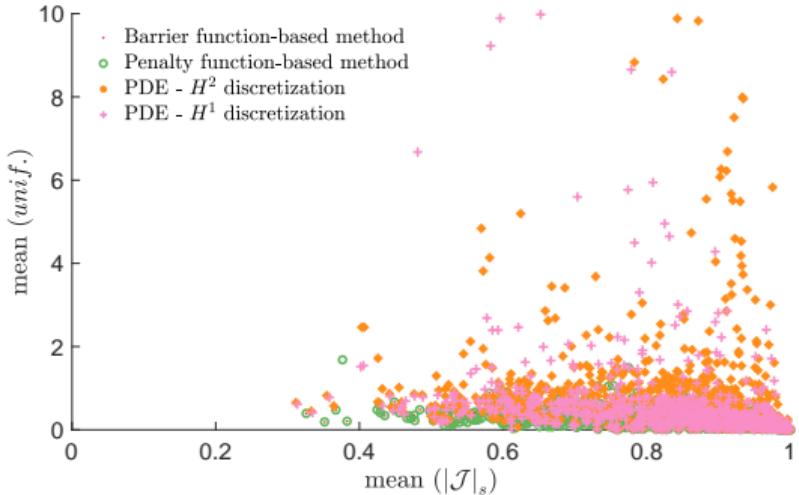
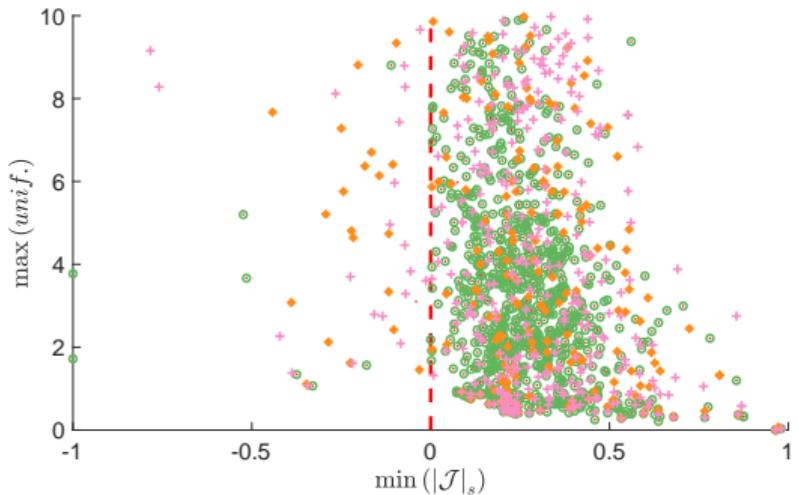
Effectiveness and Quality Assessment



Success rates:

- PDE - H^2 discretization [Hinz+2018]: $608/977 \simeq 62.23\%$;
- PDE - H^1 discretization [Ji+2023]: $721/977 \simeq 73.80\%$;

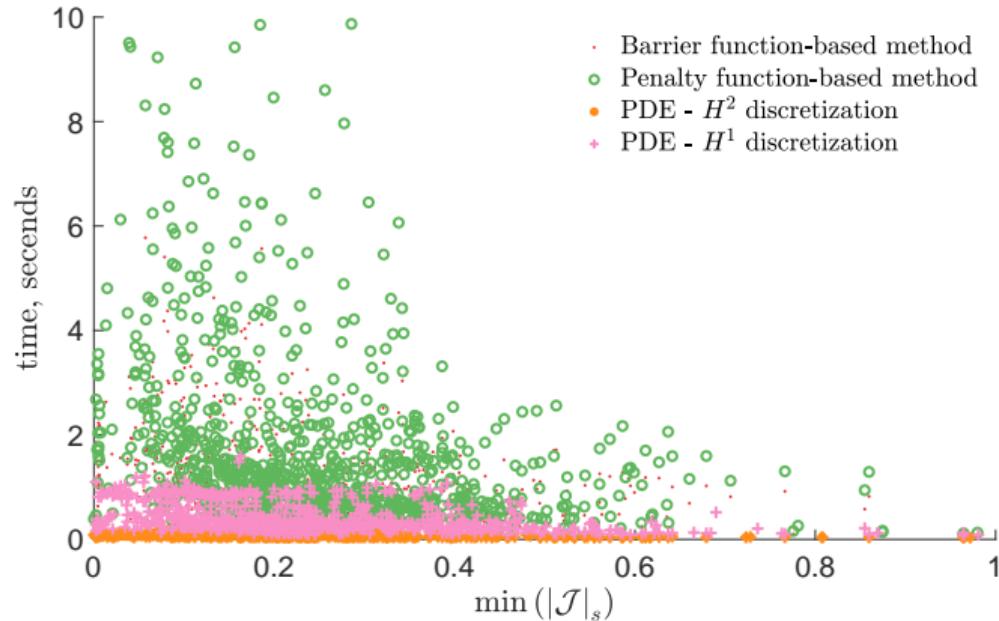
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- PDE - H^2 discretization [Hinz+2018]: $608/977 \simeq 62.23\%$;
- PDE - H^1 discretization [Ji+2023]: $721/977 \simeq 73.80\%$;
- Barrier function-based method [Ji+2021]: $961/977 \simeq \textbf{98.36\%}$;
- Penalty function-based method [Ji+2022]: $956/977 \simeq 97.85\%$.

Computation Time



- PDE-based ~ 0.2 sec., optimization based ~ 2 sec. on my laptop.
- PDE-based methods demonstrate higher efficiency.

Agenda

① Domain parameterization problem

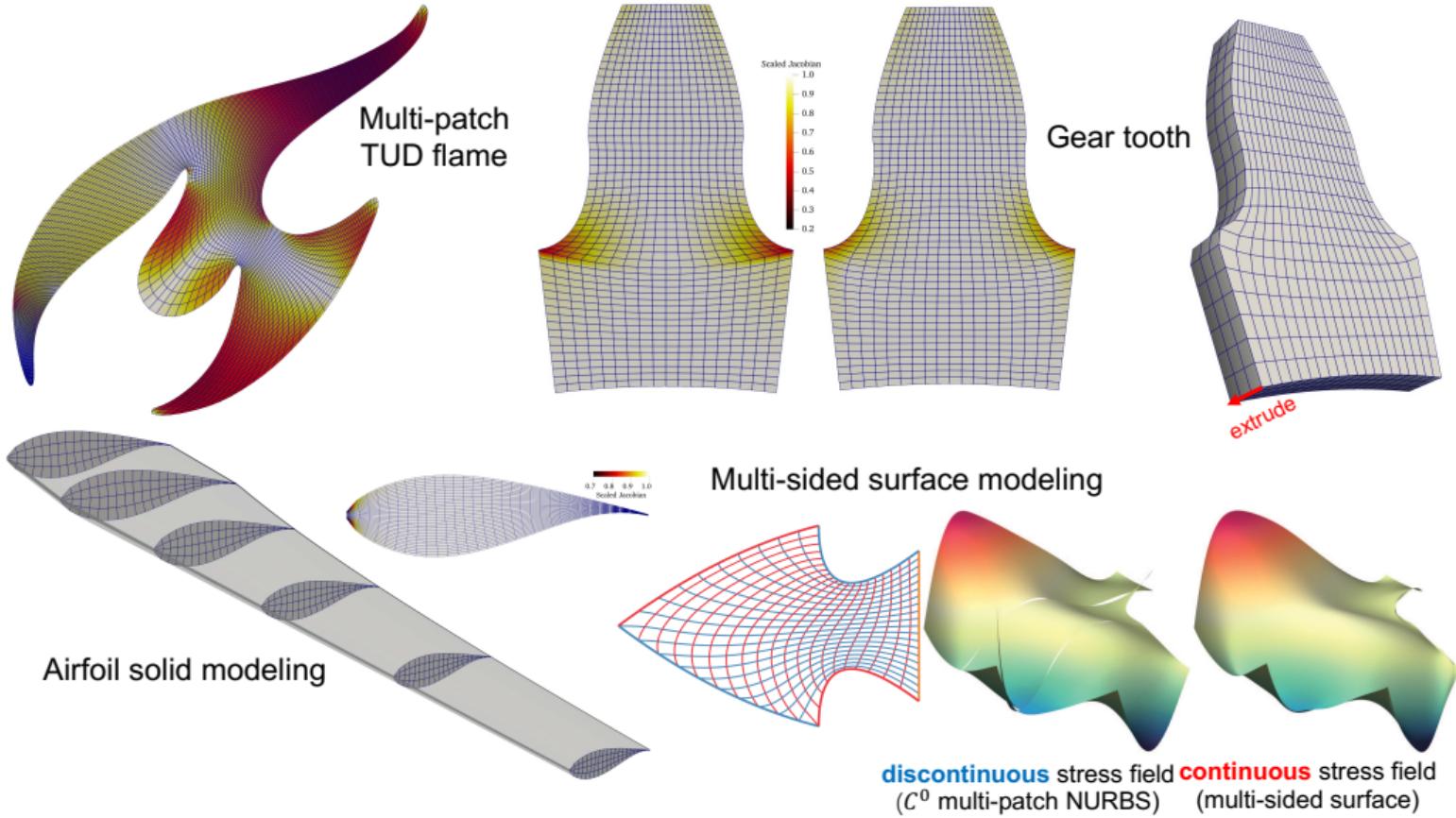
② Inside G+Smo

③ Applications

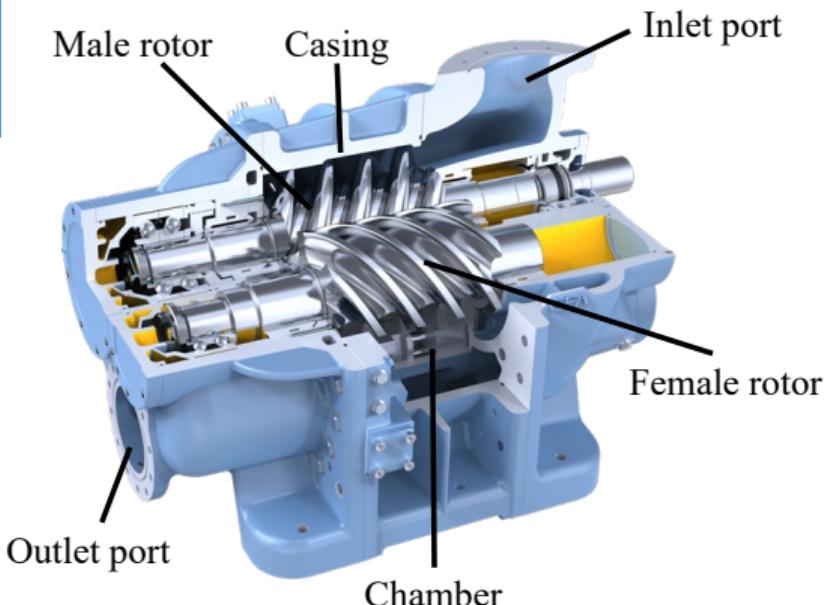
④ Outside G+Smo

⑤ Conclusions

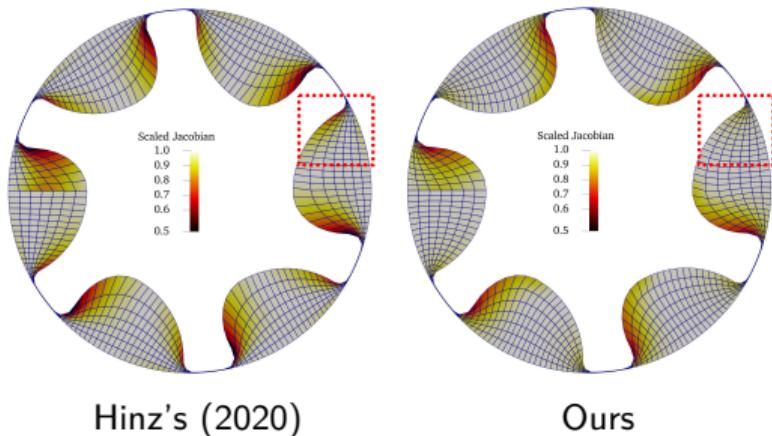
as_parameterization_example.cpp is helping people!



Rotary Twin-Screw Compressors



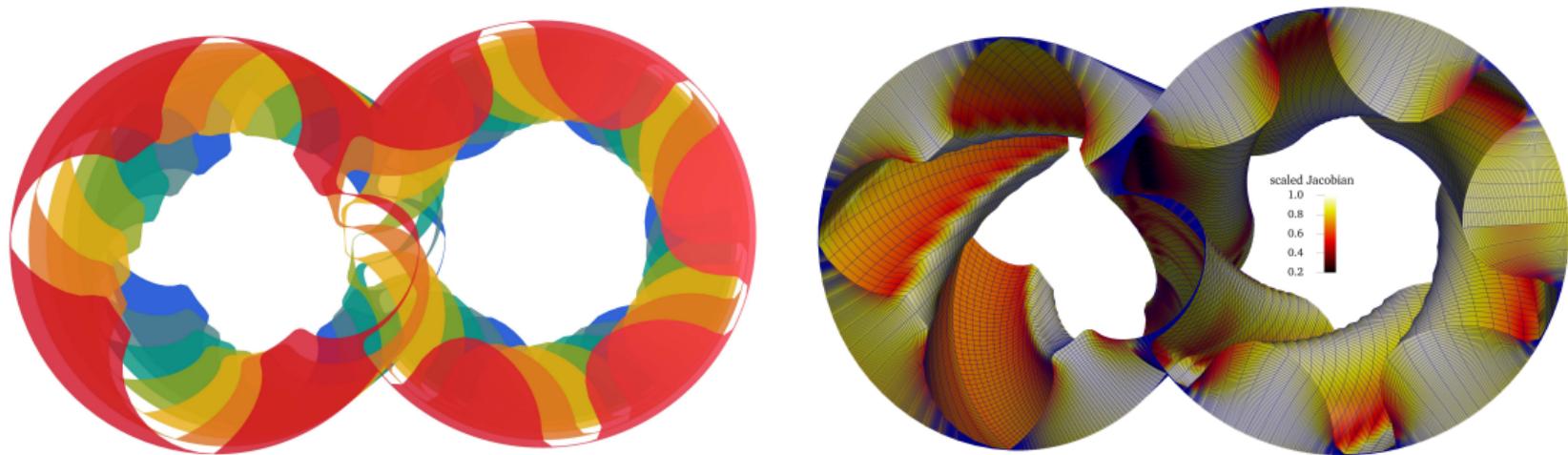
Rotary Twin-Screw Compressor (source ^a)



^a <https://www.gascompressors.co.uk/technologies/oil-flooded-screw-compressor/>

Volumetric Completion via Spline Lofting

- Complete volumetric parameterization by lofting computed slices.
- With the **PreAA** solver, achieve parameterization for twin-screw machines **in just 3 seconds** on my personal laptop.

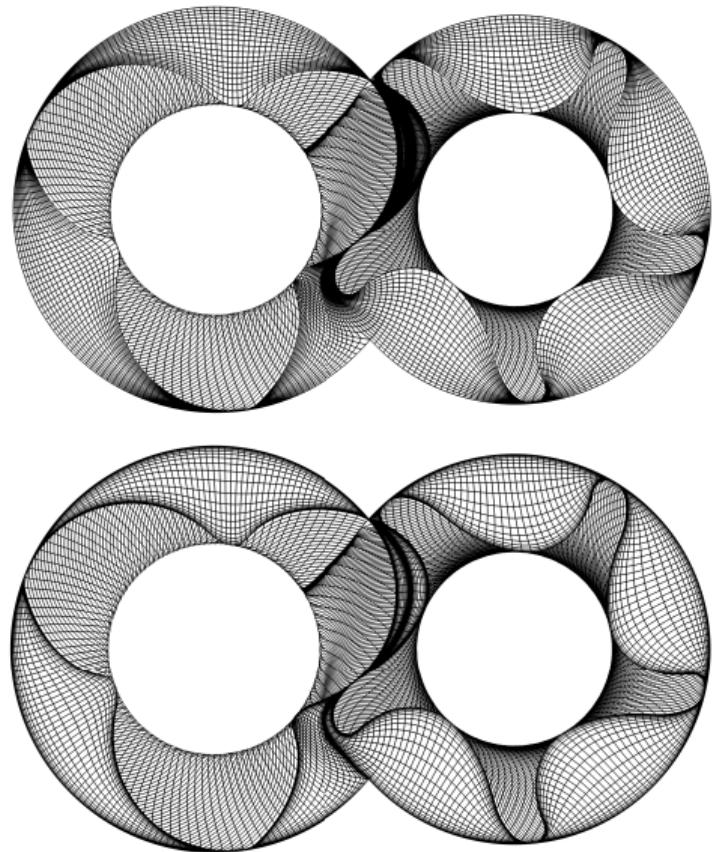


Discretization I: Boundary Layer Mesh

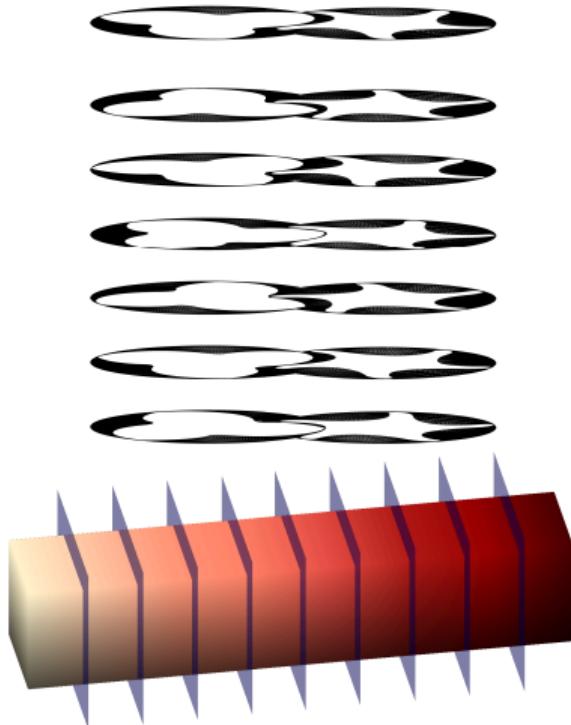
- **bsp.eval() is ALL YOU NEED.**
- Employ a simple expansion transformation:

$$\begin{cases} \xi = \hat{\xi}, \\ \eta = \frac{\tanh(\alpha(2\hat{\eta} - 1))}{2\tanh(\hat{\eta})} + \frac{1}{2}, \end{cases}$$

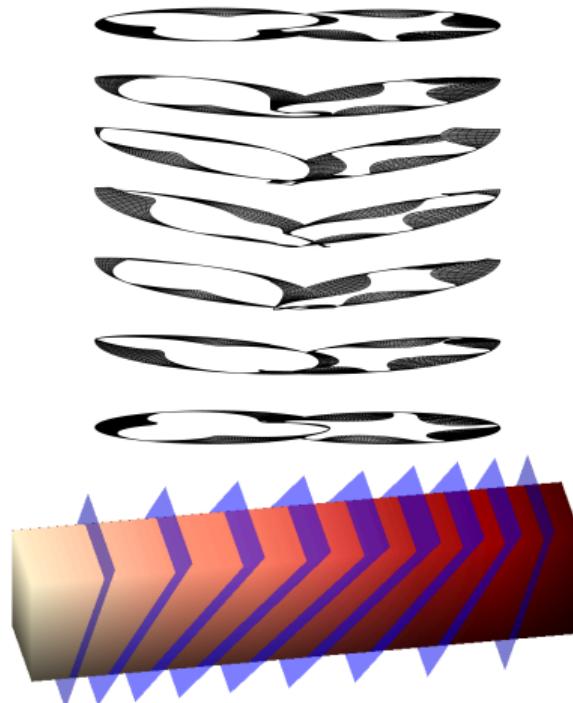
where α represents the expansion factor.



Discretization II: Flow-aligned hexahedral mesh



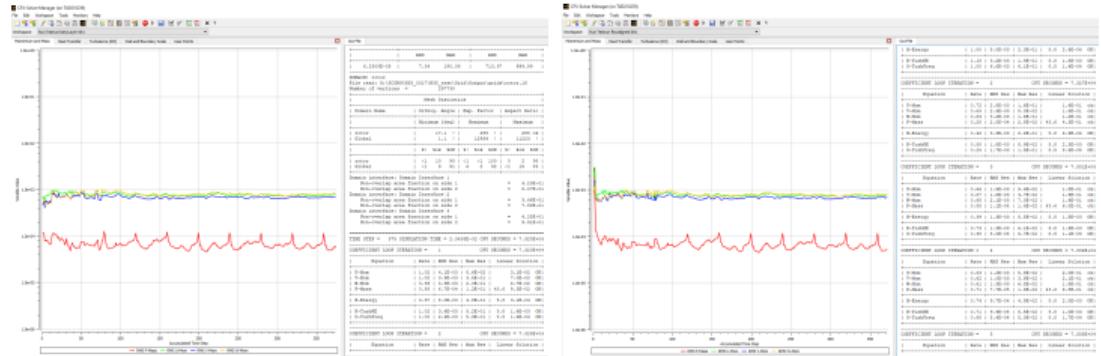
generic discretization



flow-aligned discretization

Simulation using ANASYS CFX

- Mesh density: $198 \times 95 \times 8$ for the male rotor and $200 \times 95 \times 8$ for the female rotor.



SCORG™

Boundary Layer Mesh

Flow-aligned Mesh

Agenda

① Domain parameterization problem

② Inside G+Smo

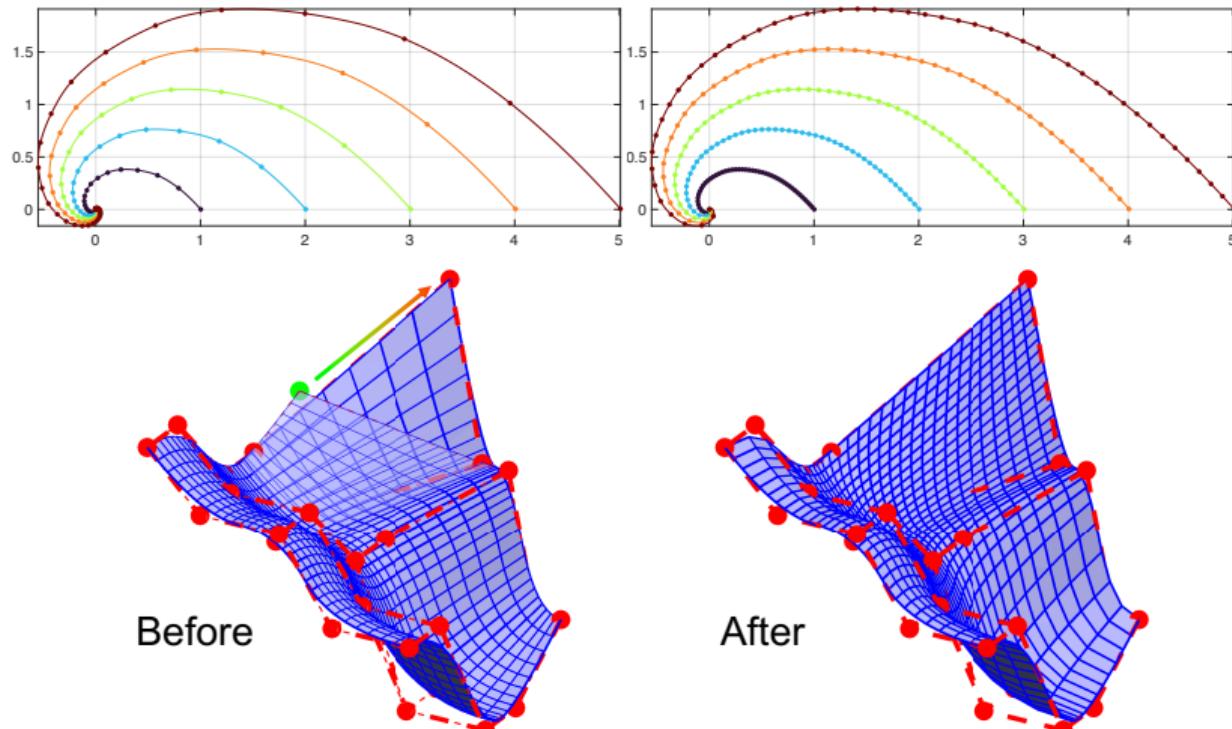
③ Applications

④ Outside G+Smo

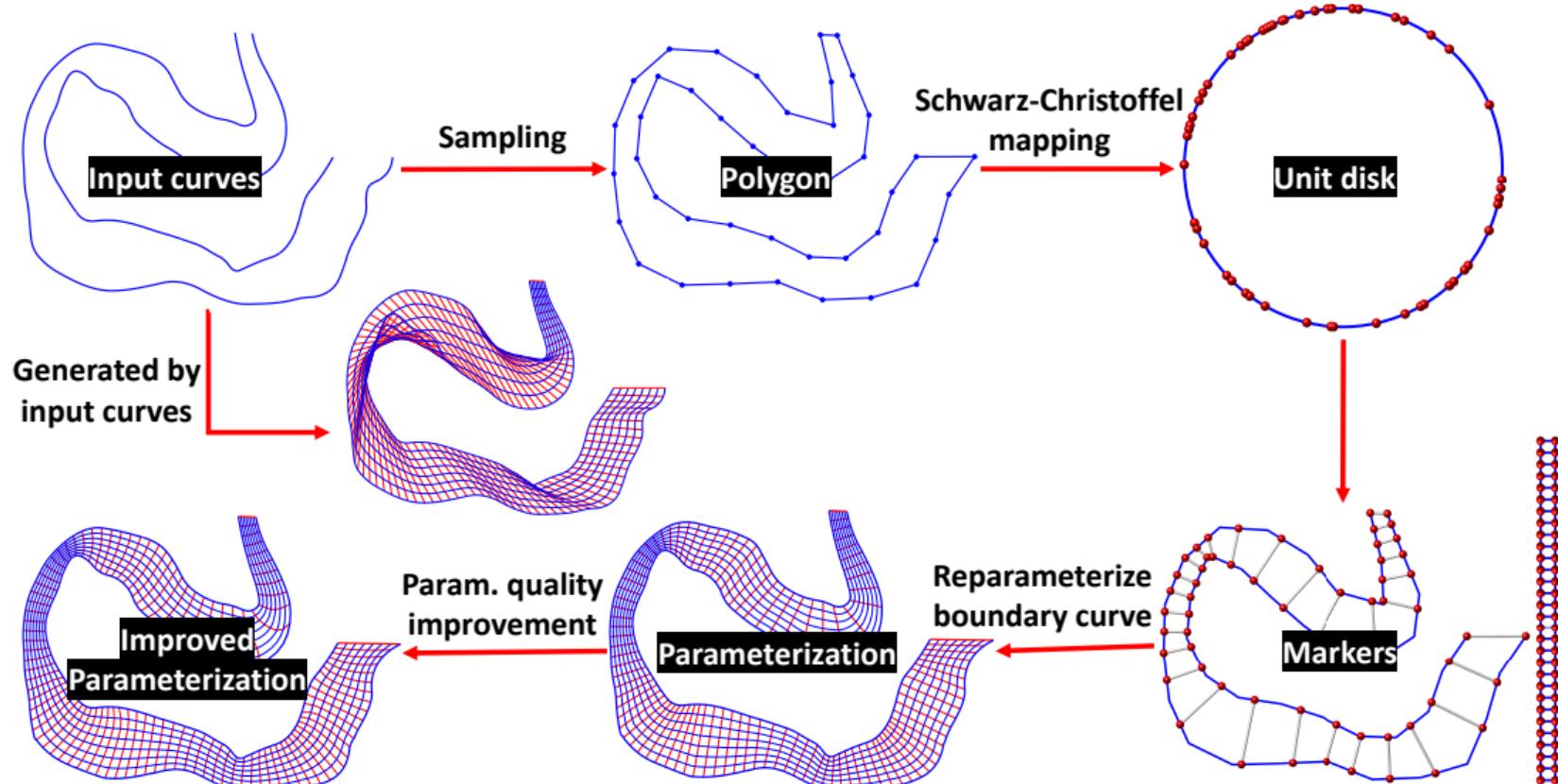
⑤ Conclusions

I: Extension to curve/surface reparameterization

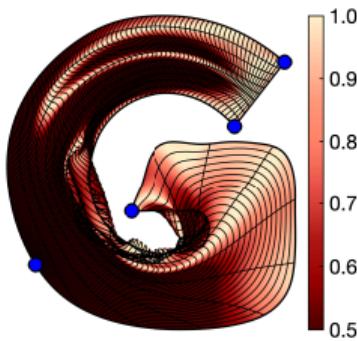
- Optimizing parameteric representation while **keeping the underlying geometry**.
- Ensures the resulting geometry **remains in NURBS form**.



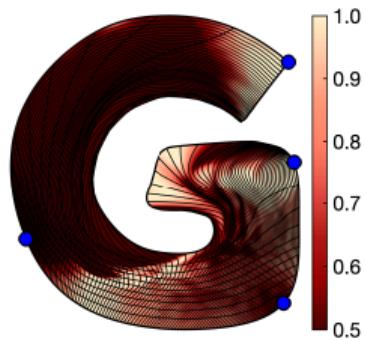
II: Parameter matching via Schwarz-Christoffel mapping



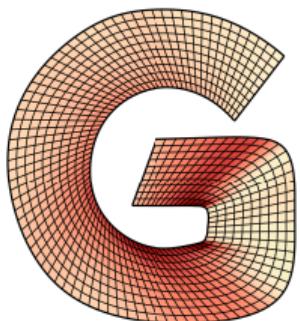
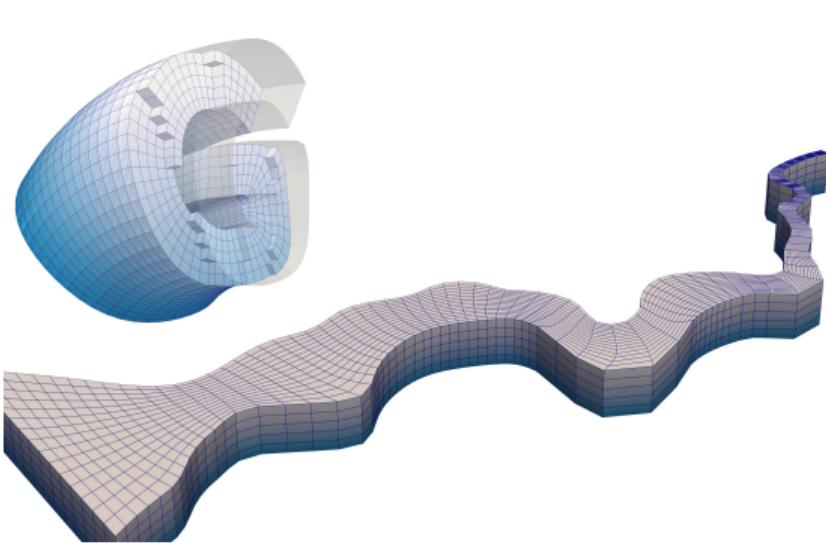
Results and Comparisons



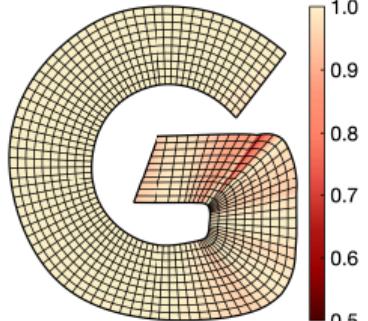
Optimal transport [1]



Deep learning [2]



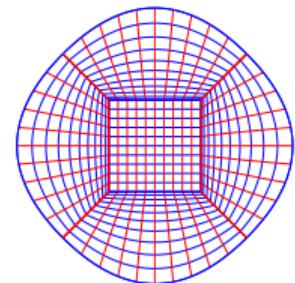
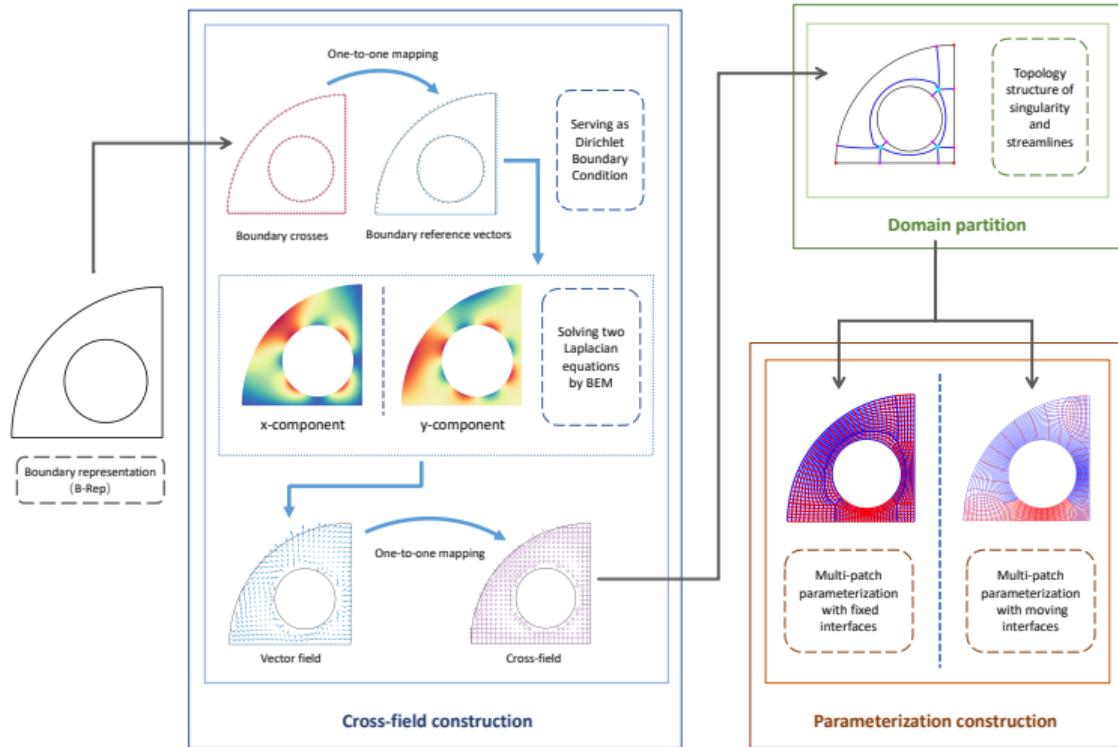
Input [3]



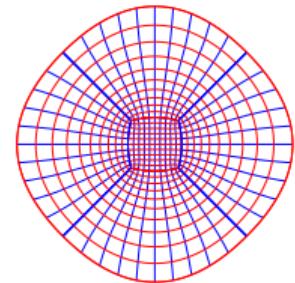
Ours [3]

- [1] Zheng, Y., Pan, M., & Chen, F. (2019). Boundary correspondence of planar domains for isogeometric analysis based on optimal mass transport. *Computer-Aided Design*, 114, 28-36.
- [2] Zhan, Z., Zheng, Y., Wang, W., & Chen, F. (2023). Boundary Correspondence for Iso-Geometric Analysis Based on Deep Learning. *Communications in Mathematics and Statistics*, 11(1), 131-150.
- [3] Ji, Y., Möller M., Yu Y., & C. Zhu Boundary parameter matching for isogeometric analysis using Schwarz-Christoffel mapping. Submitted.

III: Multi-patch Parameterization using Cross-field



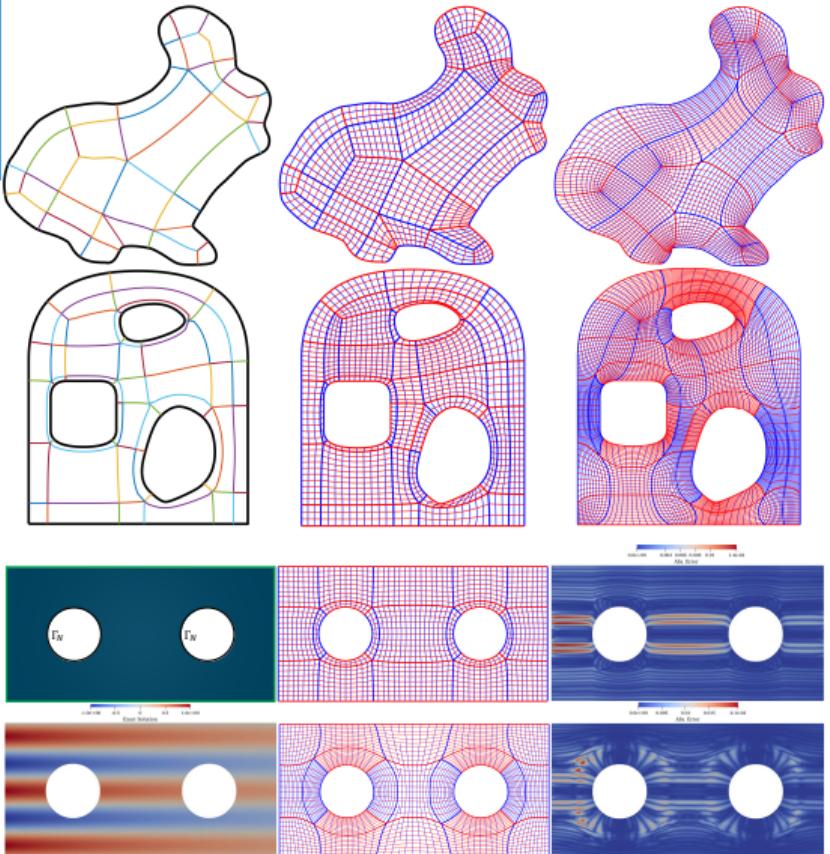
AS- G^1 [1]



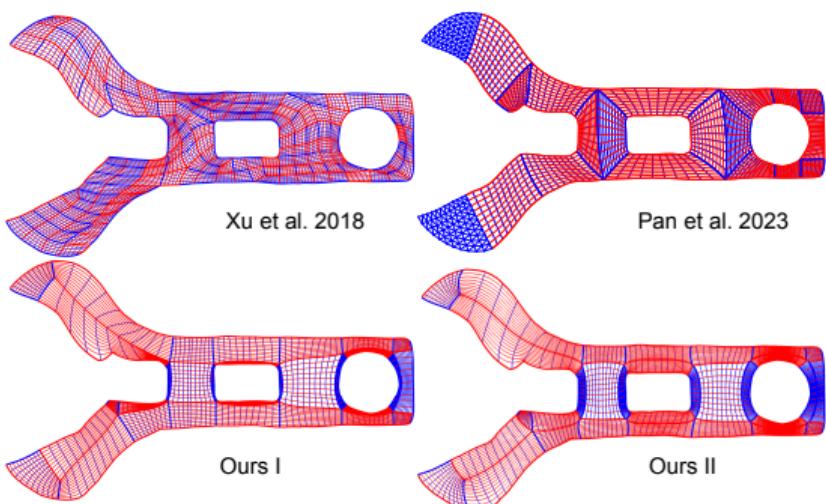
Ours [2]

- [1] Collin, A., Sangalli, G., & Takacs, T. (2016). Analysis-suitable G^1 multi-patch parametrizations ... *Computer Aided Geometric Design*, 47, 93-113.
- [2] Zhang, Y., Ji, Y., & Zhu, C. (2024). Multi-patch parameterization method for isogeometric analysis using singular structure of cross-field. *Computers and Mathematics with Applications*, accepted.

Results and Comparisons



Model	#Patch	Method	$ \mathcal{J} _s$		unif.		
			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	0.9571	0.9968



IV: r-adaptivity - Anisotropic phenomena in physics

Wave Propagation. [source](#)

Laser Printing. [source](#)

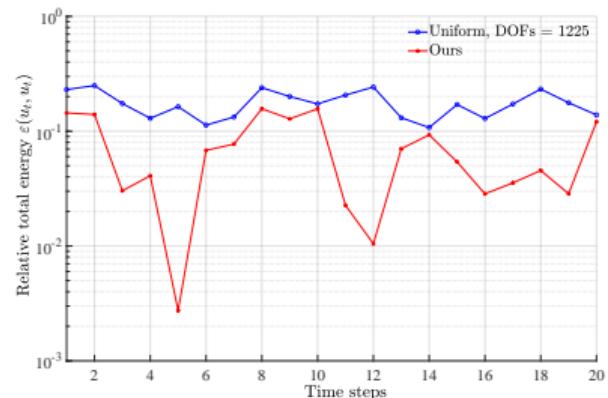
Stress Concentration. [source](#)

- **Localized and anisotropic features extensively exist** in physical phenomena;
- Isotropic parameterizations are not efficient for such problems;
- **Anisotropic parameterizations (*r*-adaptivity)**
 - Enhance per-DOF accuracy while keeping constant total DOFs.

Application to Time-dependent Dynamic PDE

- Consider a 2D **linear heat transfer problem** with a moving Gaussian heat source [1]:

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



$u(\mathbf{x}, t)$ and the parameterizations

Errors vs. time instants t

[1] Carraturo, M., Giannelli, C., Reali, A., & Vázquez, R. (2019). Suitably graded THB-spline refinement and coarsening: Towards an adaptive isogeometric analysis of additive manufacturing processes. Computer Methods in Applied Mechanics and Engineering, 348, 660-679.

Agenda

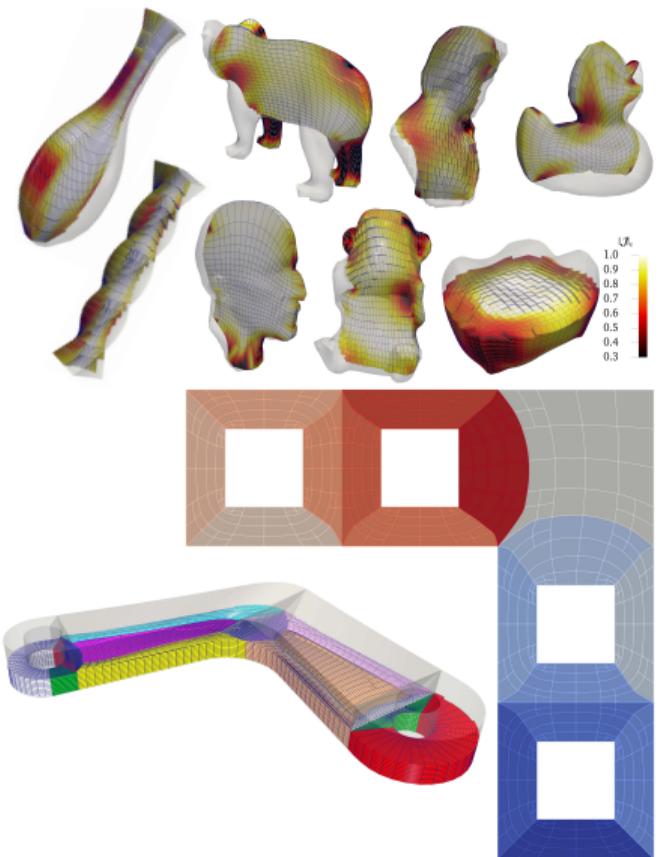
- ① Domain parameterization problem
- ② Inside G+Smo
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- ④ Outside G+Smo
- ⑤ Conclusions

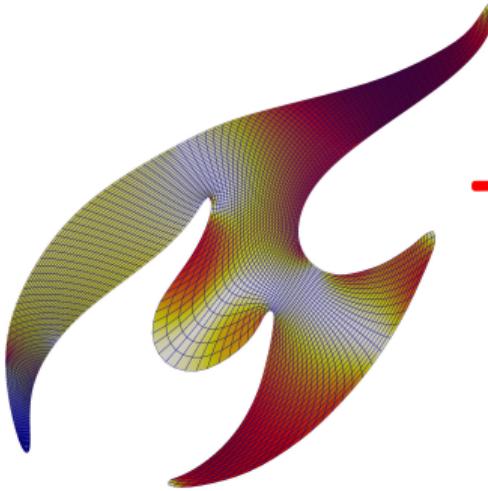
Conclusions

- G+Smo integrates **three major classes** of parameterization methods;
- Demonstrates **improved robustness and efficiency** over existing methods;
- Applicability in **real-world industry scenarios**;
- Many developed methods are still **not** in G+Smo.

Future Work:

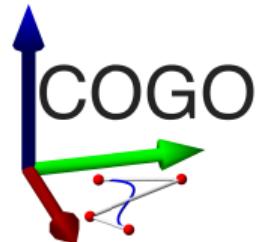
- **Incorporation of our developed and currently under-development methods** into G+Smo;
- **User-friendly** graphical user interface.





Thanks for Your Attention!

Q&A.



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

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