

# Adaptive Tangential Relaxation and Surface Fitting for High-Order Mesh Optimization

29<sup>th</sup> International Meshing Roundtable  
(virtual conference)



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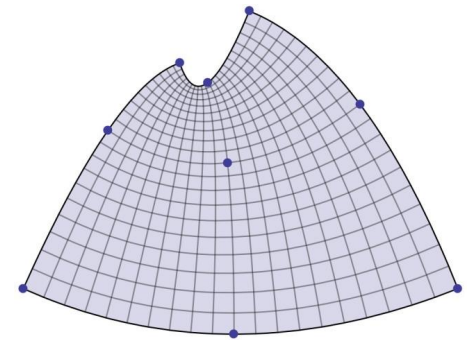


# Goal: derive mesh optimization methods based on finite element operations

- The mesh positions are represented as a FE function.

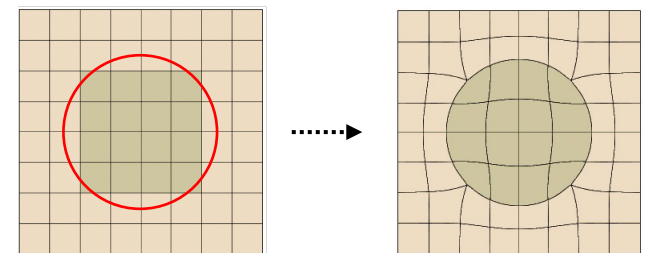
$$\mathbf{x} = (\mathbf{x}_1 \dots \mathbf{x}_N)^T, \quad x_q(\bar{x}_q) = \sum_{i=1}^N \mathbf{x}_i \bar{w}_i(\bar{x}_q)$$

$$A_q(x) = \frac{\partial x_q}{\partial \bar{x}_q} = \sum_{i=1}^N \mathbf{x}_i [\nabla \bar{w}_i(\bar{x}_q)]^T$$



*Example of a single  $Q_2$  element*

- Node movement with fixed topology.
  - Generality w.r.t. dimension and element type.
  - Avoid geometric operations.
  - Computational performance.
- Focus of this talk: extend the approach to tangential optimization and fitting.
  - Implicitly defined surfaces (level set functions).



*Example of fitting through node movement*

# Approach overview: Target-Matrix Optimization Paradigm (TMOP) and variational minimization

- Target construction: the user defines ideal target elements by specifying the target Jacobians  $W$ .

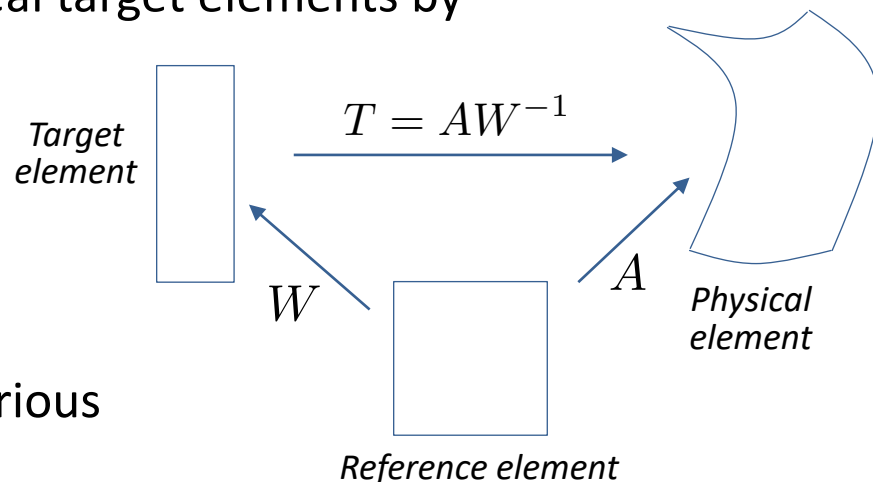
- The Jacobian  $T$  is used to define the local mesh quality measure  $\mu(T)$ .

- Combinations of  $W$  and  $\mu(T)$  control various properties of the physical elements.

$W = [\text{volume}] [\text{orientation}] [\text{skew}] [\text{aspect ratio}]$ .

- Variational minimization over the target elements (solving  $\partial F(x) / \partial x = 0$ ):

$$F(x) := \sum_{E \in \mathcal{M}} \int_{E_t} \mu(T(x_t)) dx_t = \sum_{E \in \mathcal{M}} \sum_{x_q \in E_t} w_q \det(W(\bar{x}_q)) \mu(T(x_q))$$

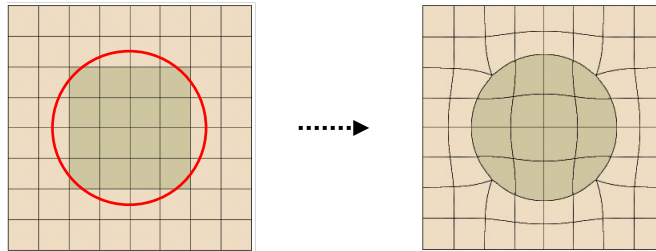


Dobrev, Knupp, Kolev, Mittal, Tomov, “The Target-Matrix Optimization Paradigm for high-order meshes”, SISC, 2019

Knupp, “Metric Type in the Target-matrix Mesh Optimization Paradigm” LLNL-TR-817490, 2020.

# Fitting and tangential relaxation are enforced weakly by a variational penalty term

- The surface of interest is given as a discrete level set (no analytic parametrization).

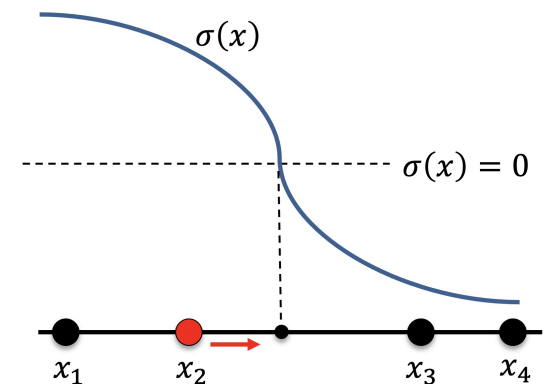
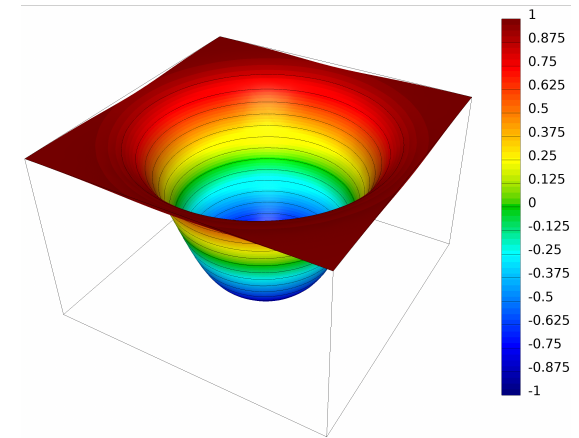


*Example of 2D interface fitting*

- Penalty formulation (quality / fitting tradeoff).
  - All mesh nodes move simultaneously.
  - One approach for fitting / tangential relaxation.

$$F(x) = F_\mu + w_\sigma \int_\Omega \bar{\sigma}(x)^2$$

- The restricted level set function  $\bar{\sigma}$  penalizes the deviation from the zero level set.
  - Marking is not a trivial procedure.



*The extra term affects only the position of the red DOF*

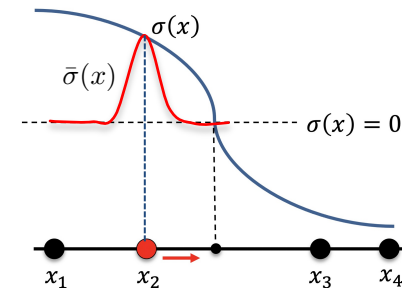


# The penalty term is put in a form that can be easily differentiated

- We restrict  $\sigma$  to the set of marked nodes:

$$\bar{\sigma}_i = \begin{cases} \sigma_i & \text{if } i \in \mathcal{S}, \\ 0 & \text{otherwise.} \end{cases}$$

- Goal: move the mesh so that  $\sigma_i \equiv \sigma(x_i) = 0$ .
  - Interpolatory finite element basis functions.
  - The nodes of  $x$  and  $\sigma$  must coincide.



- The derivative computation must see the term  $\partial \sigma_i \backslash \partial x \neq 0$ .

$$\bar{\sigma}(\mathbf{x}) = \sum_{i \in \mathcal{S}} \sigma_i \phi_i(\mathbf{x}) = \sum_{s \in \mathcal{S}} \sum_{k=1}^{N_x} \sigma_k \phi_k(\mathbf{x}_i) \phi_i(\mathbf{x}).$$

$$\phi_k(\mathbf{x}_s) = \delta_{ks}$$

*Derivatives of this term represent the changes of  $\sigma$  as the mesh evolves*

- Field remap – required as  $\sigma$  is a discrete FE function.

Dobrev, Knupp, Kolev, Tomov, “Towards simulation-driven optimization of high-order meshes by TMOP”, IMR, 2019.

# Derivatives of the penalization term are calculated through standard FE operations

$$\begin{aligned}\frac{\partial F_\sigma}{\partial x_{a,i}} &= \frac{2\omega_\sigma}{c_\sigma} \int_{E_t} \bar{\sigma}(\mathbf{x}) \frac{\partial \bar{\sigma}(\mathbf{x})}{\partial x_a} \frac{\partial x_a}{\partial x_{a,i}} \\ &= \frac{2\omega_\sigma}{c_\sigma} \int_{E_t} \bar{\sigma}(\mathbf{x}) \sum_{s \in \mathcal{S}} \sum_k \sigma_k \left( \frac{\partial \phi_k(\mathbf{x}_s)}{\partial x_a} \phi_s(\mathbf{x}) + \right. \\ &\quad \left. \phi_k(\mathbf{x}_s) \frac{\partial \phi_s(\mathbf{x})}{\partial x_a} \right) w_i(\bar{\mathbf{x}}),\end{aligned}$$

$$\begin{aligned}\frac{\partial^2 F_\sigma}{\partial x_{b,j} \partial x_{a,i}} &= \frac{2\omega_\sigma}{c_\sigma} \int_{E_t} \left( \frac{\partial \bar{\sigma}(\mathbf{x})}{\partial x_b} \frac{\partial \bar{\sigma}(\mathbf{x})}{\partial x_a} + \right. \\ &\quad \left. \bar{\sigma}(\mathbf{x}) \frac{\partial^2 \bar{\sigma}(\mathbf{x})}{\partial x_b \partial x_a} \right) \frac{\partial x_a}{\partial x_{a,i}} \frac{\partial x_b}{\partial x_{b,j}} \\ &= \frac{2\omega_\sigma}{c_\sigma} \int_{E_t} (\mathcal{D}_a \mathcal{D}_b + \bar{\sigma}(\mathbf{x}) \mathcal{D}^2) w_i(\bar{\mathbf{x}}) w_j(\bar{\mathbf{x}}),\end{aligned}$$

where

$$\begin{aligned}\mathcal{D}_* &= \sum_{s \in \mathcal{S}} \sum_k \sigma_k \left( \frac{\partial \phi_k(\mathbf{x}_s)}{\partial x_*} \phi_s(\mathbf{x}) + \phi_k(\mathbf{x}_s) \frac{\partial \phi_s(\mathbf{x})}{\partial x_*} \right), \\ \mathcal{D}^2 &= \sum_{s \in \mathcal{S}} \sum_k \sigma_k \left( \frac{\partial \phi_k(\mathbf{x}_s)}{\partial x_a} \frac{\partial \phi_s(\mathbf{x})}{\partial x_b} + \frac{\partial^2 \phi_k(\mathbf{x}_s)}{\partial x_b \partial x_a} \phi_s(\mathbf{x}) + \right. \\ &\quad \left. \frac{\partial \phi_k(\mathbf{x}_s)}{\partial x_b} \frac{\partial \phi_s(\mathbf{x})}{\partial x_a} + \phi_k(\mathbf{x}_s) \frac{\partial^2 \phi_s(\mathbf{x})}{\partial x_b \partial x_a} \right),\end{aligned}$$

$$a, b = 1 \dots d, \quad i, j = 1 \dots N_x.$$

Knupp, Kolev, Mittal, Tomov, “Adaptive surface fitting and tangential relaxation for high-order mesh optimization”, IMR, 2021.

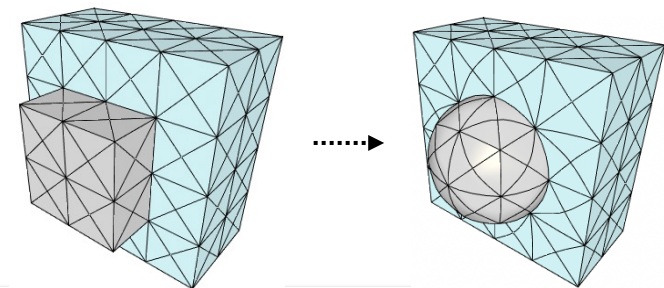
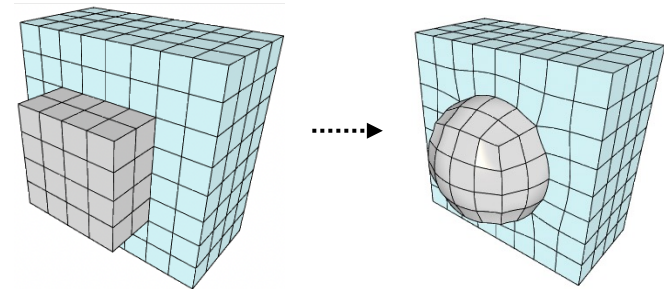
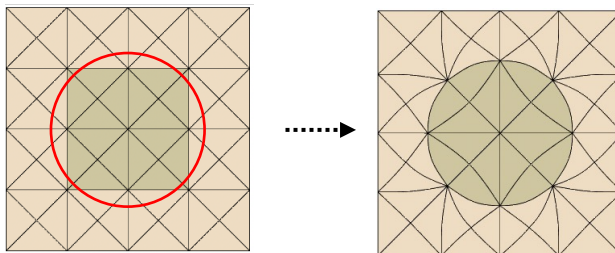
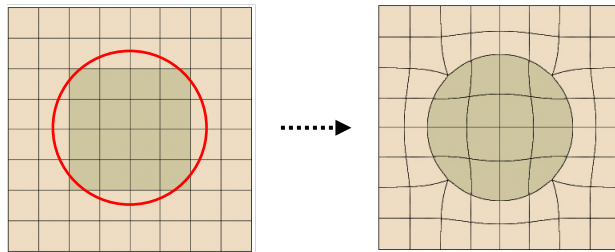
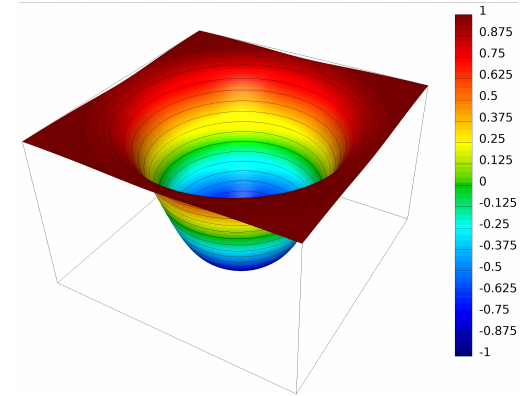
# The method behaves as expected on academic problems with smooth interfaces

- In all cases  $\sigma$  is a discrete FE function.
  - Ball at the center of the domain.

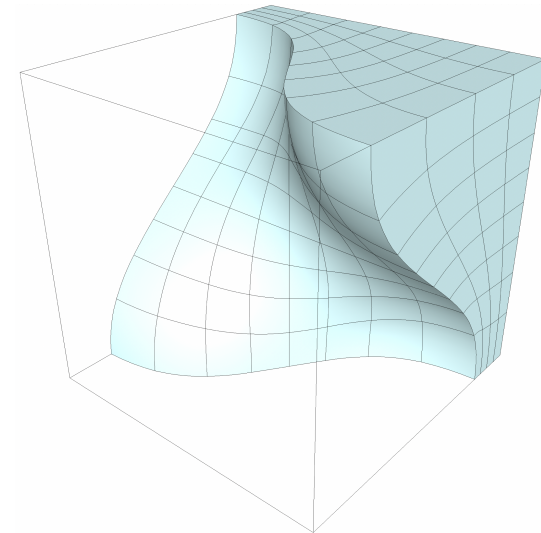
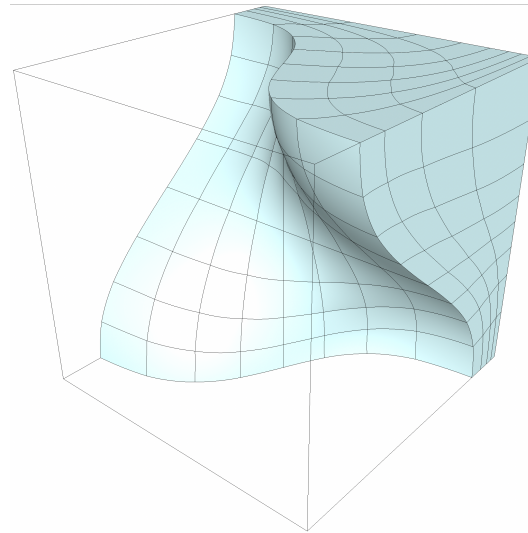
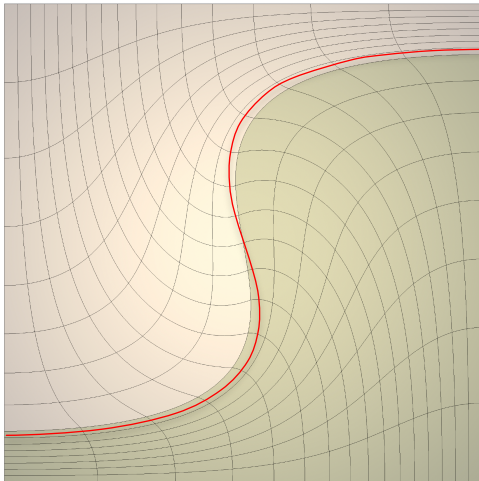
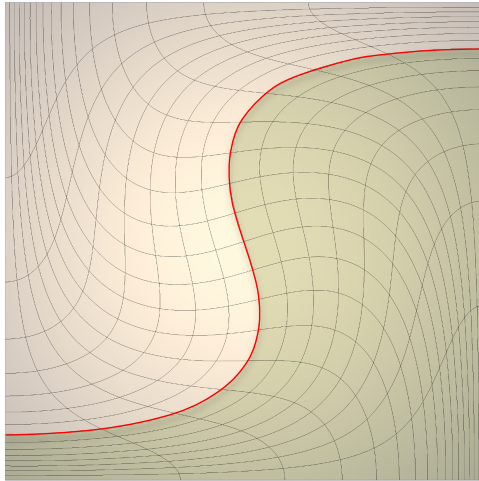
$$2D \text{ metric: } \mu_{80} = (1 - \gamma) \left( 0.5 \frac{|T|^2}{\tau} - 1 \right) + \gamma \left( 0.5 \left( \tau - \frac{1}{\tau} \right)^2 \right)$$

$$3D \text{ metric: } \mu_{333} = (1 - \gamma) \left( \frac{|T|^2 |T^{-1}|^2}{9} - 1 \right) + \gamma \left( 0.5 \left( \tau + \frac{1}{\tau} \right) - 1 \right)$$

P. Knupp, "*Metric Type in the Target-matrix Mesh Optimization Paradigm*" LLNL-TR-817490, 2020.



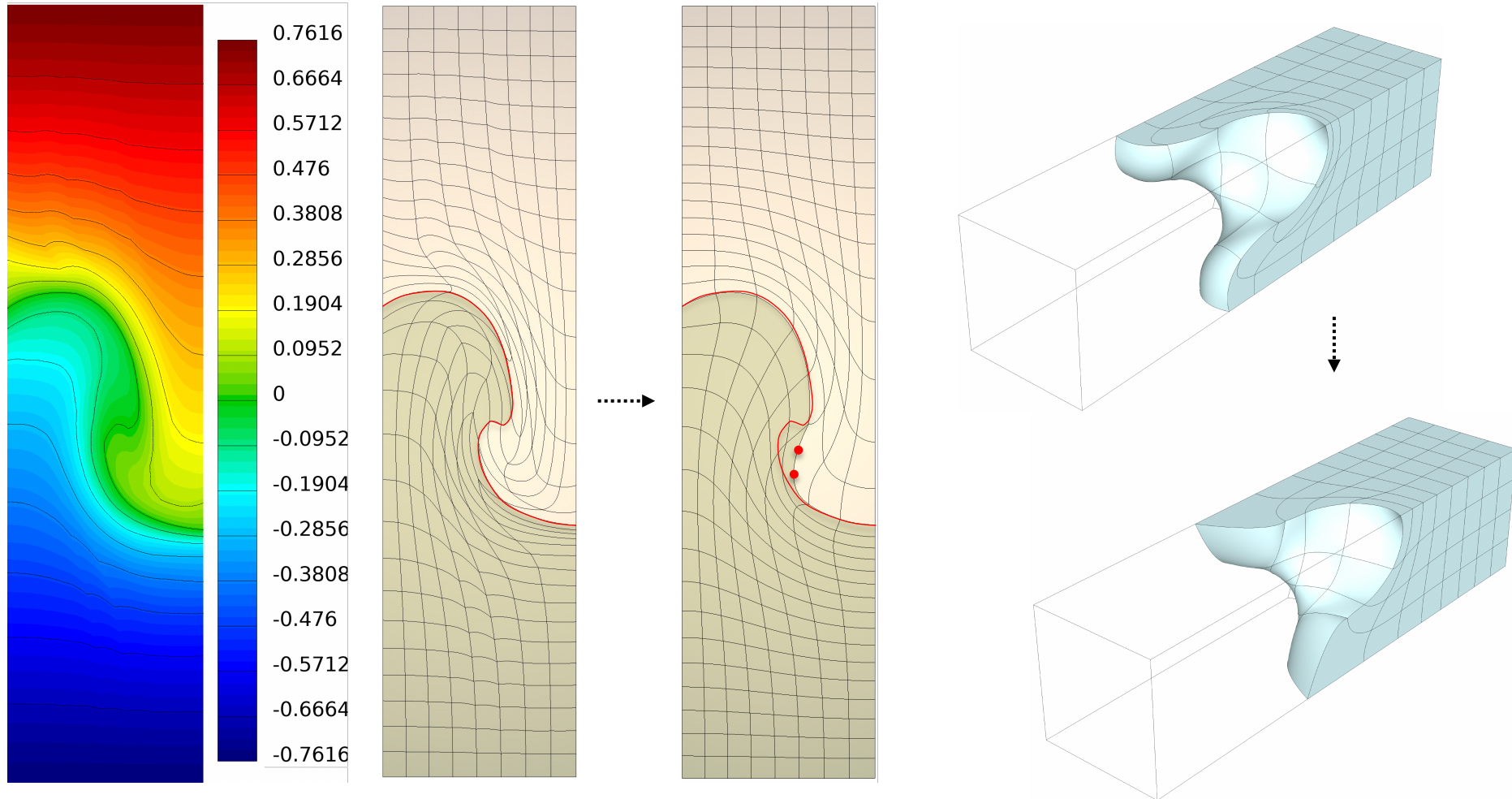
# The method behaves as expected on academic problems with smooth interfaces



Approach	$F$ decrease	$\mathcal{E}_{avg}$	$\mathcal{E}_{max}$
Fixed interface	34.4%	0	0
$w_\sigma = 250$	51.4%	8.2e-2	1.3e-1
$w_\sigma = 1000$	42.6%	3.6e-2	6.4e-2

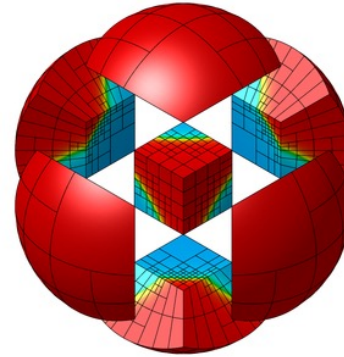


# For non-smooth practical cases, further research and methods are required

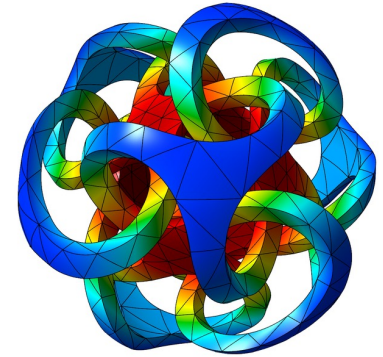


# MFEM open source implementation

- All presented methods are (or will be) available in MFEM.
- MFEM contains **12** 2D metrics, **7** 3D metrics, all metric derivatives, **6** target construction methods.
- User interface provided by the ***mesh\_optimizer*** and ***pmesh\_optimizer*** miniapps.
  - Choice of target construction / quality metric / adaptivity fields / parameters.
  - Visualization through GLVis.



*mfem.org*



*glvis.org*

Dobrev, Knupp, Kolev, Mittal, Tomov, “*The Target-Matrix Optimization Paradigm for high-order meshes*”, SIAM J. Sci. Comp., 2019.

Dobrev, Knupp, Kolev, Mittal, Rieben, Tomov, “*Simulation-driven optimization of high-order meshes in ALE hydrodynamics*”, Comput. Fluids, 2020.

Knupp, Kolev, Mittal, Tomov, “*Adaptive surface fitting and tangential relaxation for high-order mesh optimization*”, IMR, 2021.

# Summary and future work

- Surface fitting and tangential relaxation through an adaptive FE formulation.
  - Discrete representation of the surface; no analytic parametrization.
  - Exploratory attempt to use FE-only operations.
- Weak enforcement through a variational penalty term.
  - Simultaneous optimization of surface and non-surface nodes.
  - Applicable to both fitting and tangential relaxation.
- Derivatives are computed through standard FE operations.
- Behaves as expected on smooth surfaces.
- Future work – robust behavior for practical non-smooth problems.



# CASC

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