

Computing multiple solutions of topology optimization problems

USNCCM17



John Papadopoulos¹



Patrick Farrell²



Thomas Surowiec³

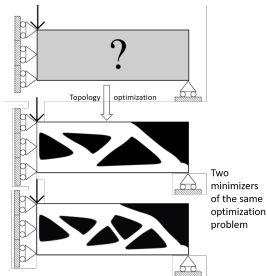


Endre Süli²

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Objective

Find the optimal distribution of a continuum that minimizes a problem-specific cost functional with no prior knowledge of the optimal shape or topology.



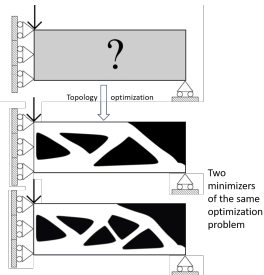
Aage, Andreassen, Lazarov, Sigmund, *Nature* (2017)

Models for topology optimization problems tend to:

- involve PDEs \implies require a discretization;
- be nonconvex \implies may support multiple local minima.

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Choice of optimization strategy

Observations

- Potentially many (local) minimizers.
- Millions of degrees of freedom.

Consequences

- Require quickly converging algorithms.
- Compute multiple minimizers in a systematic manner.
- Require preconditioners for the solves e.g. effective multigrid cycles.

Our proposal

The deflated barrier method.

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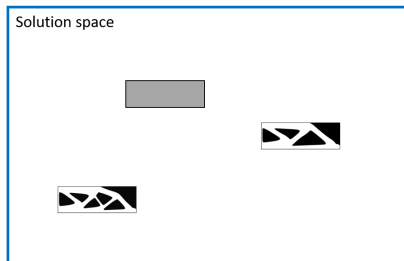
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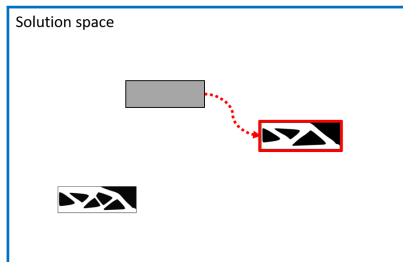
Barrier-like terms + primal-dual active set strategy + deflation



The deflated barrier method

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Barrier-like terms + primal-dual active set strategy + deflation

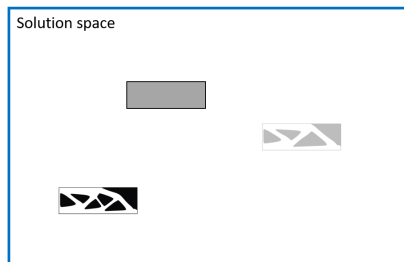


Step I: optimize from initial guess

The deflated barrier method

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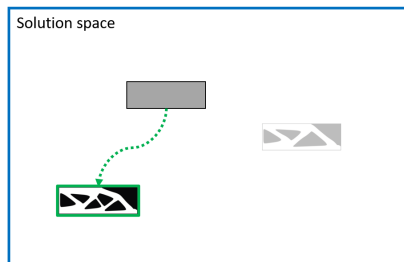


Step II: deflate solution found

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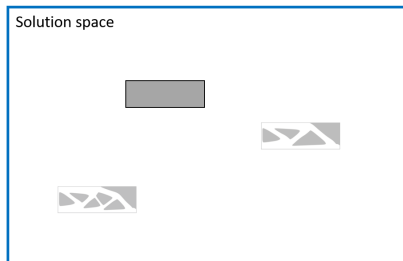


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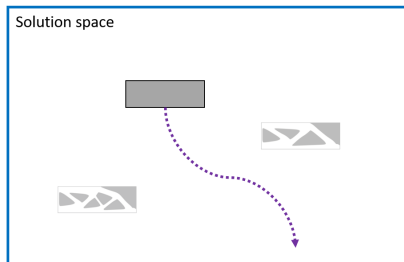


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Step III: termination on nonconvergence

Construction of deflated problems

A nonlinear transformation of first-order optimality conditions

$$\mathcal{F}(z) = 0 \rightarrow \mathcal{G}(z) := \mathcal{M}(z; r)\mathcal{F}(z) = 0.$$

A deflation operator

We say that $\mathcal{M}(z; r)$ is a deflation operator if for any sequence $z \rightarrow r$

$$\liminf_{z \rightarrow r} \|\mathcal{G}(z)\| = \liminf_{z \rightarrow r} \|\mathcal{M}(z; r)\mathcal{F}(z)\| > 0.$$

Theorem

This is a deflation operator for $p \geq 1$:

$$\mathcal{M}(z; r) = \left(\frac{1}{\|z - r\|^p} + 1 \right).$$

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Deflation is *very* easy!

Step 1

Compute the normal *undeflated* Newton update δz .

Step 2

Let $m = m(z_k) = \mathcal{M}(z_k, r)$. Then the *deflated* Newton update is

$$\delta z_D = \tau(z_k, \delta z) \delta z$$

where

$$\tau(z_k, \delta z) := \left(1 + \frac{m^{-1}(m')(\delta z)}{1 - m^{-1}(m')(\delta z)} \right).$$

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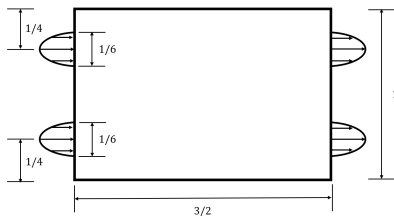
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The Borrvall–Pettersson problem

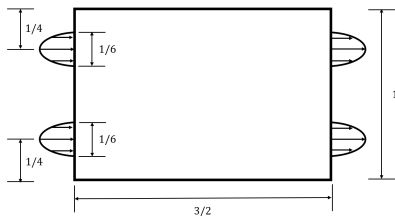


Double-pipe problem

A fluid topology optimization problem

- Stokes flow.
- Wish to minimize the power dissipation of the flow;
- Catch! The channels can occupy up to $1/3$ area.
- Requires solving a nonconvex optimization problem with PDE, box, and volume constraints.

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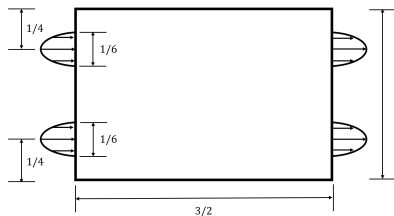


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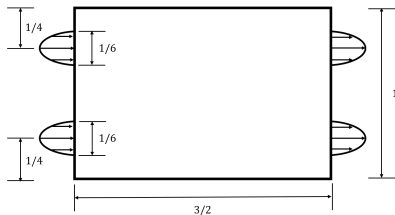


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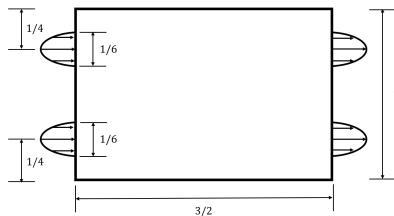


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Double-pipe solutions



(a) Straight channels

Double-pipe solutions



(a) Straight channels



(b) Double-ended wrench

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(c) Neumann (i)

Double-pipe solutions



(a) Straight channels



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(d) Neumann (ii)

Problem

Find velocity u and material distribution ρ that minimize

$$J(u, \rho) = \frac{1}{2} \int_{\Omega} \alpha(\rho) |u|^2 + |\nabla u|^2 - 2f \cdot u \, dx,$$

subject to $\operatorname{div}(u) = 0$, $0 \leq \rho \leq 1$, and $\int_{\Omega} \rho \, dx \leq \gamma |\Omega|$.

Deflated barrier method

For $\mu = \mu_0$ ($\mu \rightarrow 0$), solve $\nabla L_{\mu}(u, \rho, p, \lambda) \stackrel{\text{“=”}}{=} 0$ with a primal-dual active set strategy where

$$\begin{aligned} L_{\mu}(u, \rho, p, \lambda) = & J(u, \rho) - \int_{\Omega} p \operatorname{div}(u) + \lambda(\gamma - \rho) \, dx \\ & - \mu \int_{\Omega} \log((\rho + \epsilon)(1 + \epsilon - \rho)) \, dx. \end{aligned}$$

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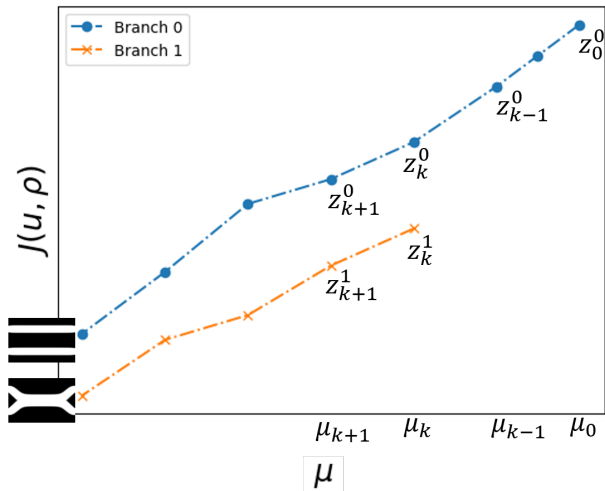
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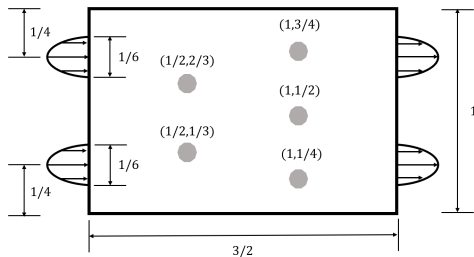
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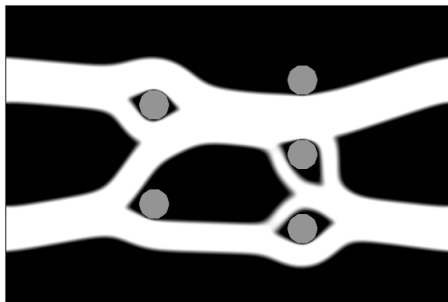


Five-holes double-pipe setup.

Fluid topology optimization

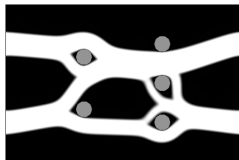
- Navier–Stokes flow.
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A fluid topology optimization problem

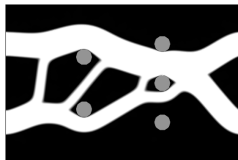


$$J = 60.09$$

A fluid topology optimization problem

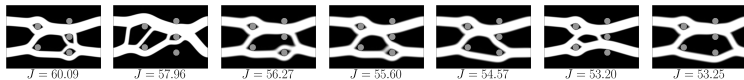


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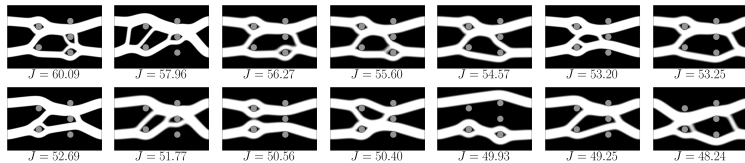


$J = 57.96$

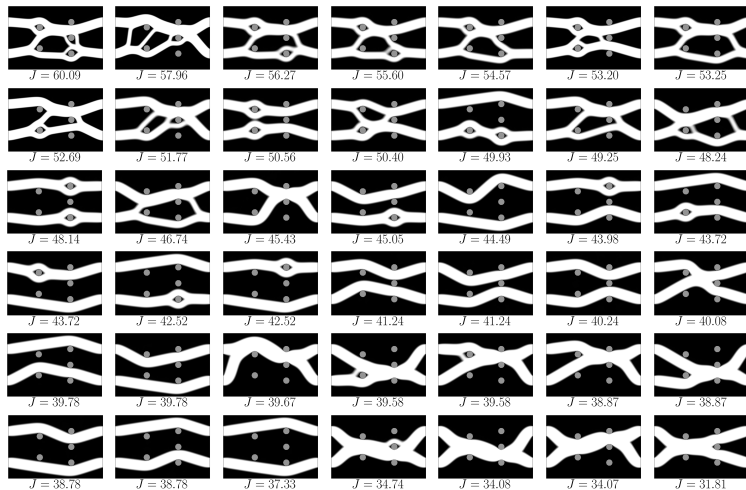
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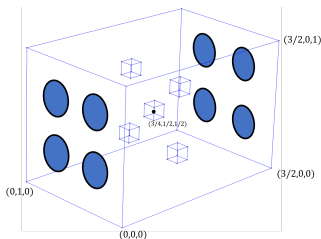


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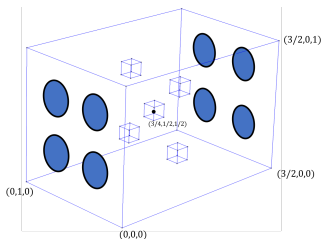
Examples

- 3D discretization on a $40 \times 40 \times 40$ block $\sim 3,000,000$ dofs.
- (Stokes) Nevertheless still numerically tractable via preconditioning techniques implemented with Firedrake 
 - Nested block preconditioning via Schur complements;
 - Augmented Lagrangian control of the pressure Schur complement;
 - Vertex-star patch type relaxation for multigrid schemes.



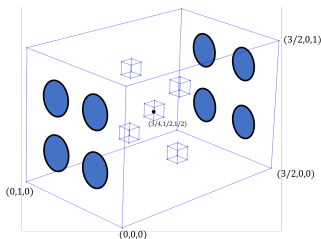
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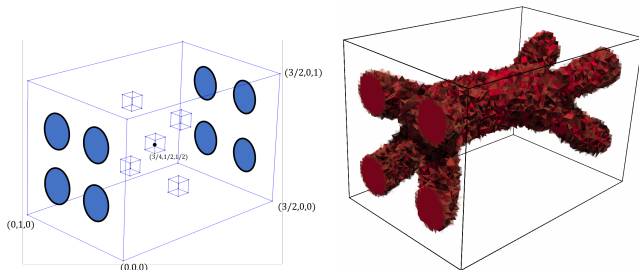
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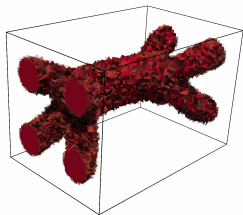


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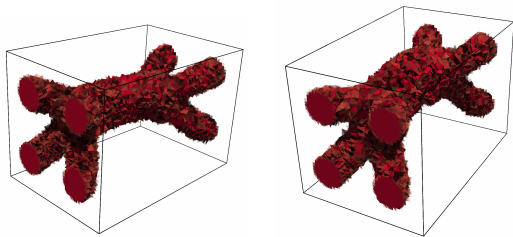
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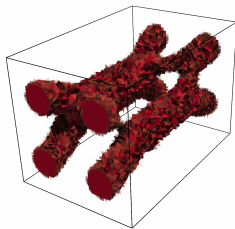
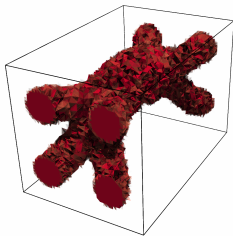
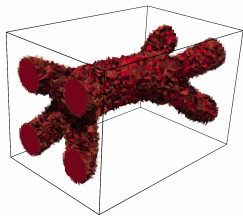
3D five-holes quadruple-pipe



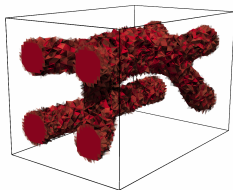
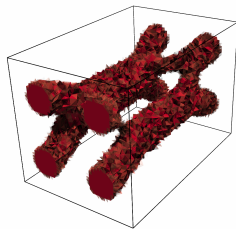
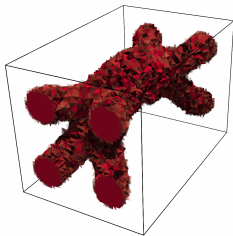
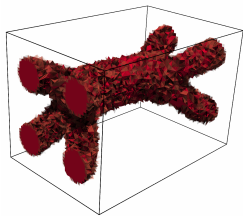
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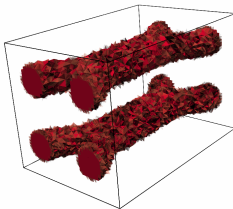
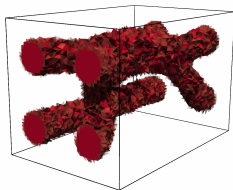
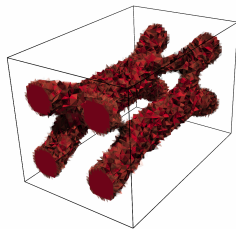
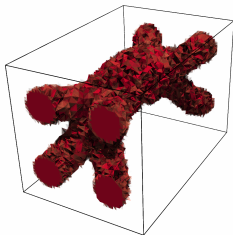
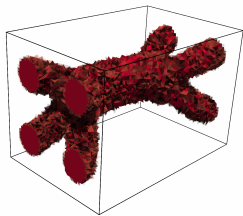
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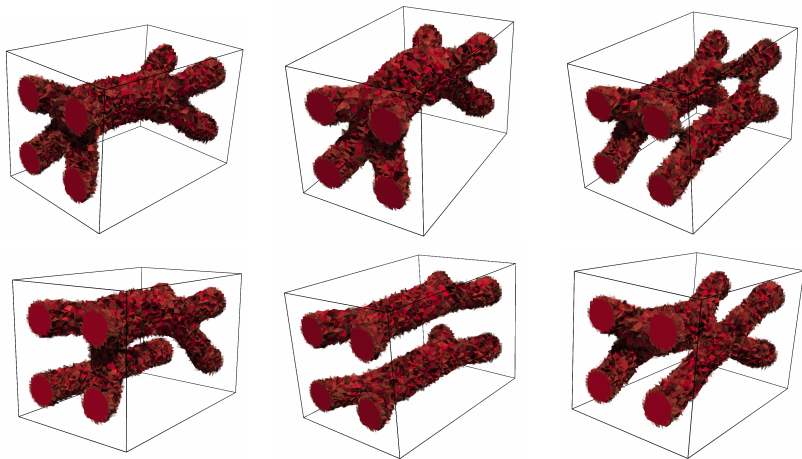
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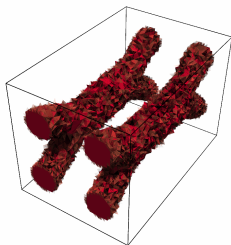
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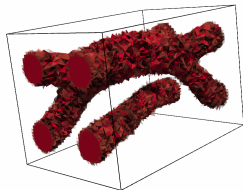
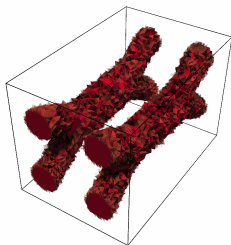
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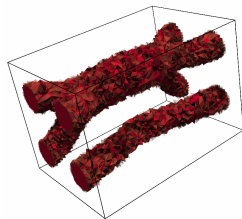
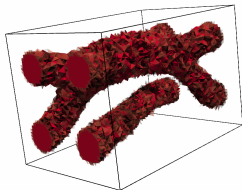
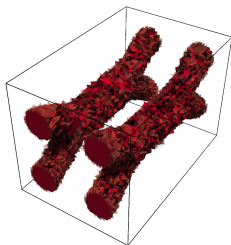
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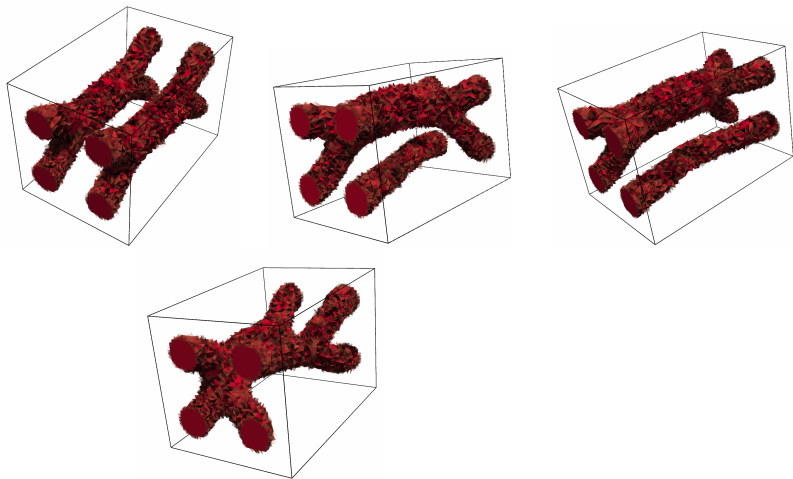
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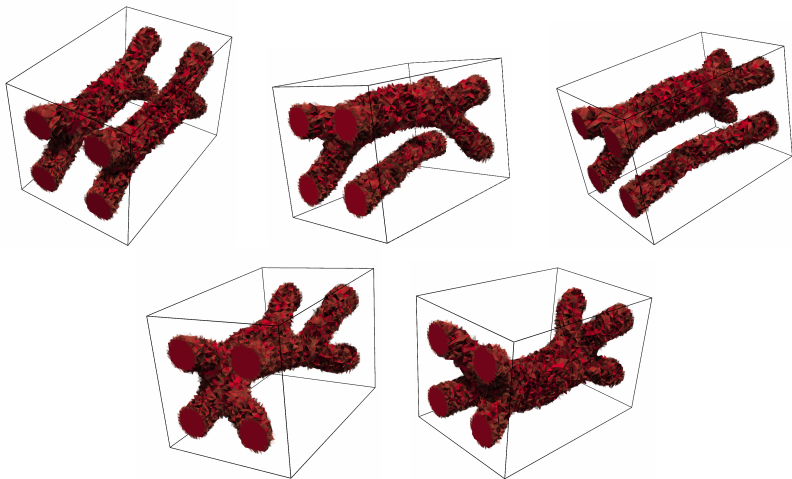
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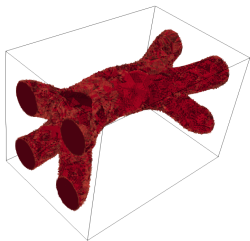
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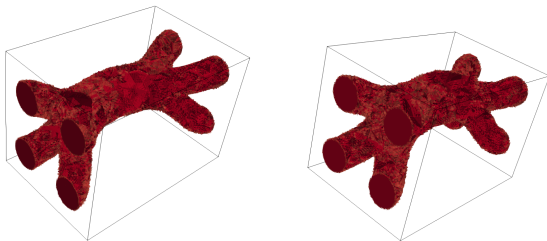


Further refinement



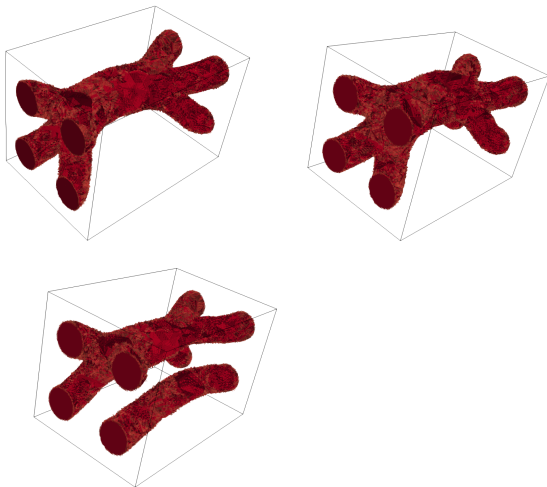
15,953,537 degrees of freedom.

Further refinement



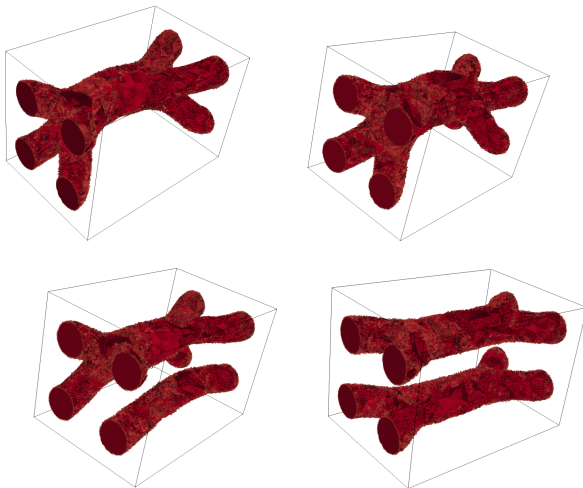
15,953,537 degrees of freedom.

Further refinement



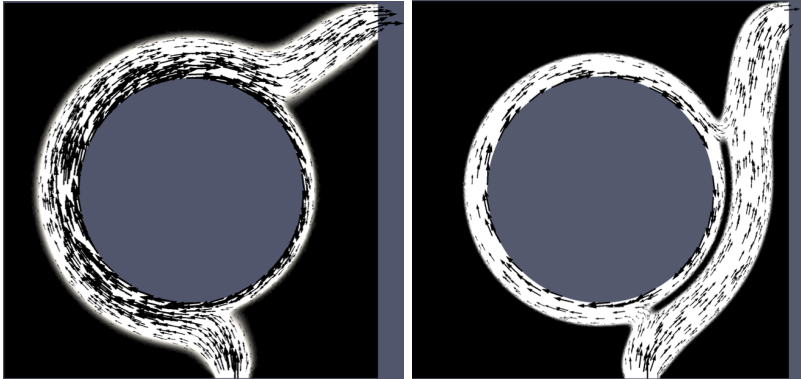
15,953,537 degrees of freedom.

Further refinement



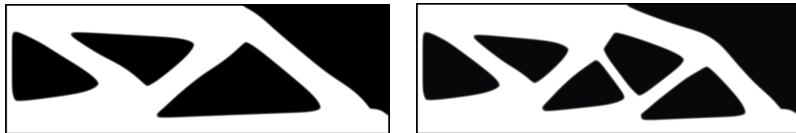
15,953,537 degrees of freedom.

More examples



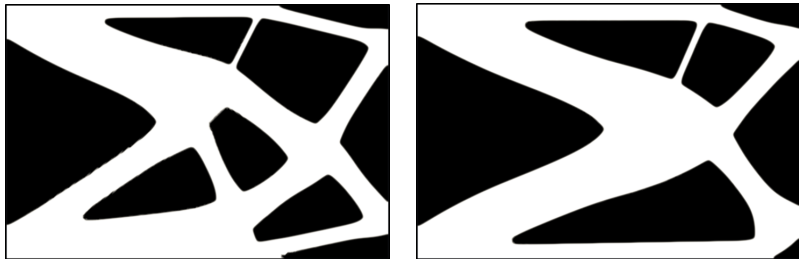
Roller pump

More examples



MBB beam

More examples



Double cantilever

Conclusions

- A strategy for computing multiple solutions of topology optimization problems.
- Barrier-like terms + active set strategy + deflation.
- Can solve large 3D problems with good preconditioners.

Computing multiple solutions of topology optimization problems

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Deflated barrier method

<https://github.com/ioannisPApapadopoulos/fir3dab>.

Deflation

<https://github.com/ioannisPApapadopoulos/Deflation>.

Deflation for bifurcation diagrams

<https://bitbucket.org/pefarrell/defcon>.

Thank you for listening!

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