

## Free-Shape Optimization of VHF Air-Core Inductors using a Constraint-Aware Genetic Algorithm

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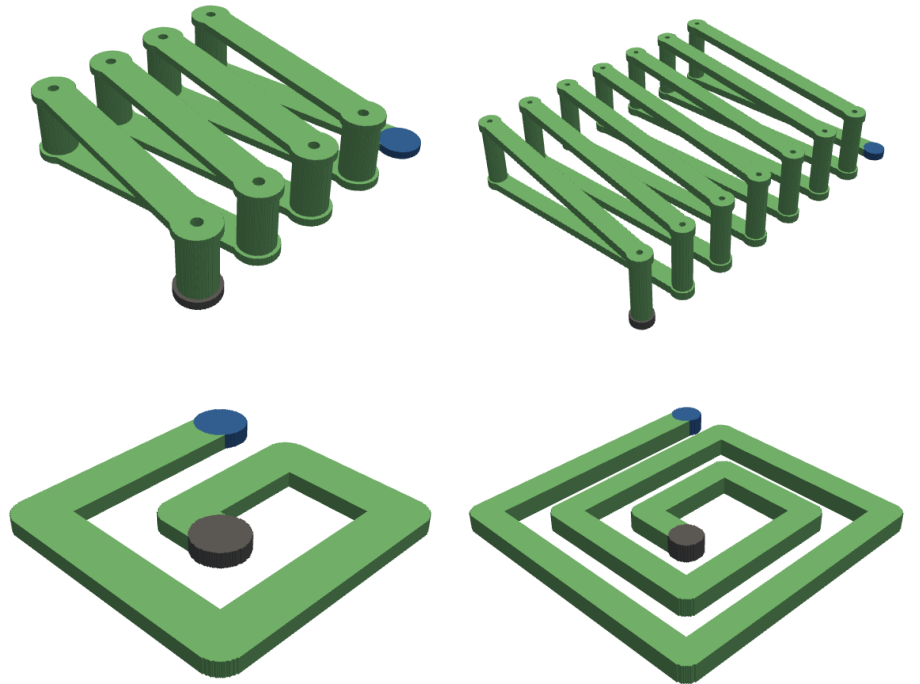


Power Management  
Integration Center



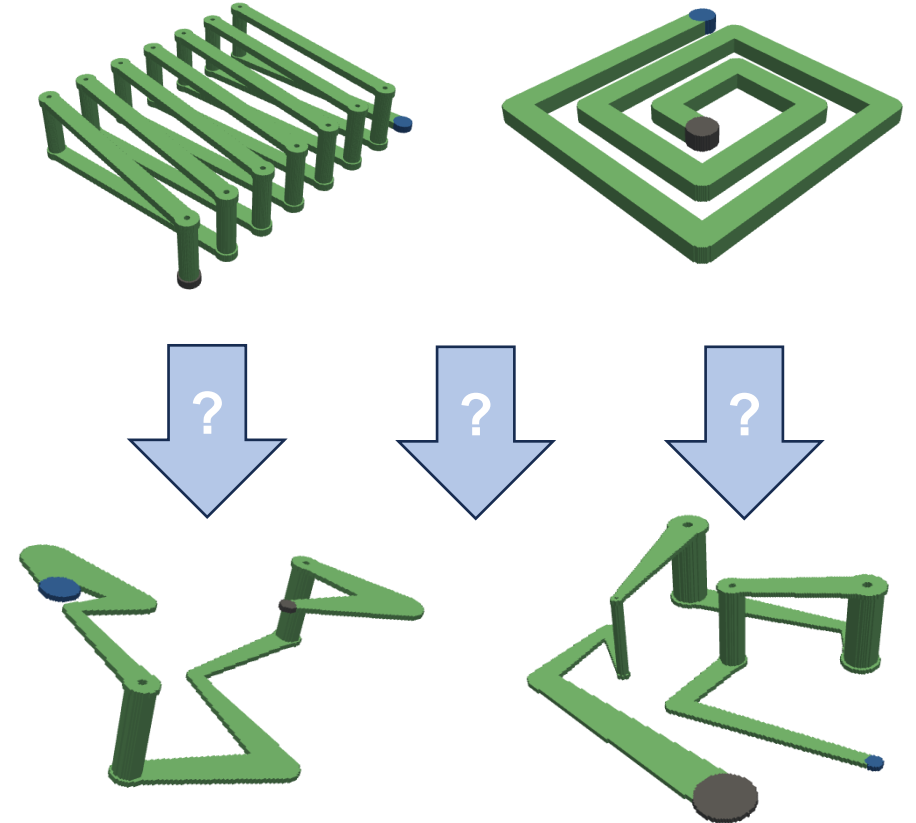
# Optimization of Magnetics

- **Magnetics** are often a **bottleneck**
- **Optimizing the materials?**
  - Copper, silver, aluminum, gold, etc.
  - Magnetic material improvements are slow
- **Optimizing the geometry?**
- **Optimal air-core inductor geometries?**
  - Spiral, solenoid, staple
  - Toroid, cylinder



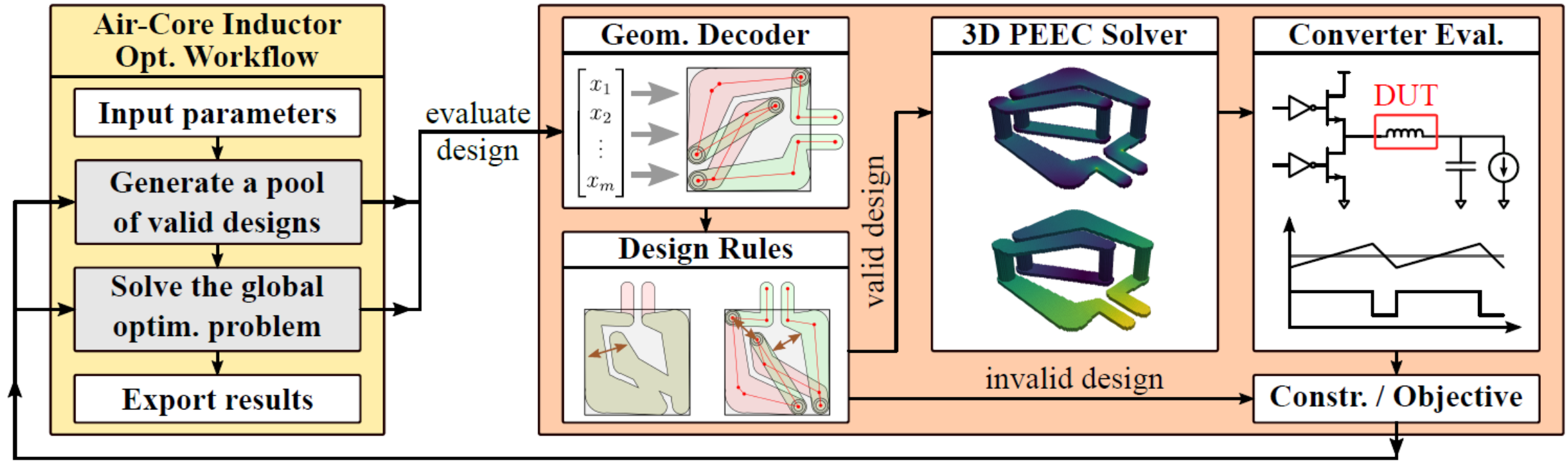
# Free-Shape Optimization

- **Topological optimization**
  - Definition of **constraints**
  - Definition of an **objective**
  - Automated shape optimization
- **Common for structural mechanics**
- **Applicability to magnetics**
  - Large computational cost
  - Complex topologies and constraints
  - Discrete variables (e.g. layers)
  - Non-differentiable / discontinuous objective



# Optimization Workflow

# Air-Core Inductor Optimization Workflow



# Geometry and Constraints

- **Considered geometries**
  - Multilayer planar process
  - Variable trace width
  - Variable via size
- **Design rules**
  - Clearance, width, footprint, angle, curvature, etc.
  - Complex computational geometry
- **Optimal designs** are often located at the **limit**
- **Optimizers** are very good at **finding loopholes**

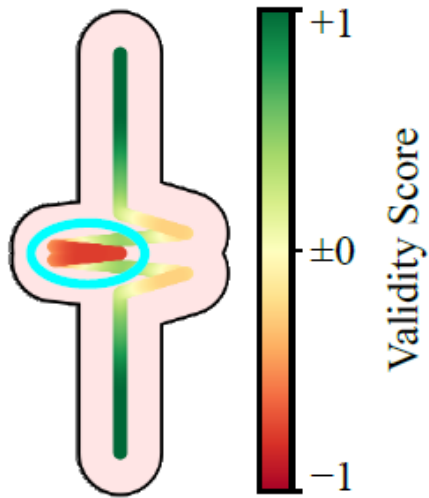
Invalid Geometries



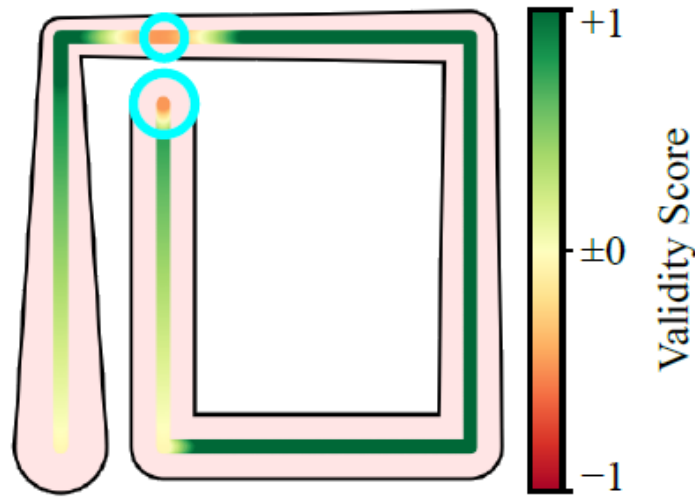
# Soft Constraints

- How “close” is a design to become valid/invalid?
- Helpful for the optimizer (as constraints and/or penalty)

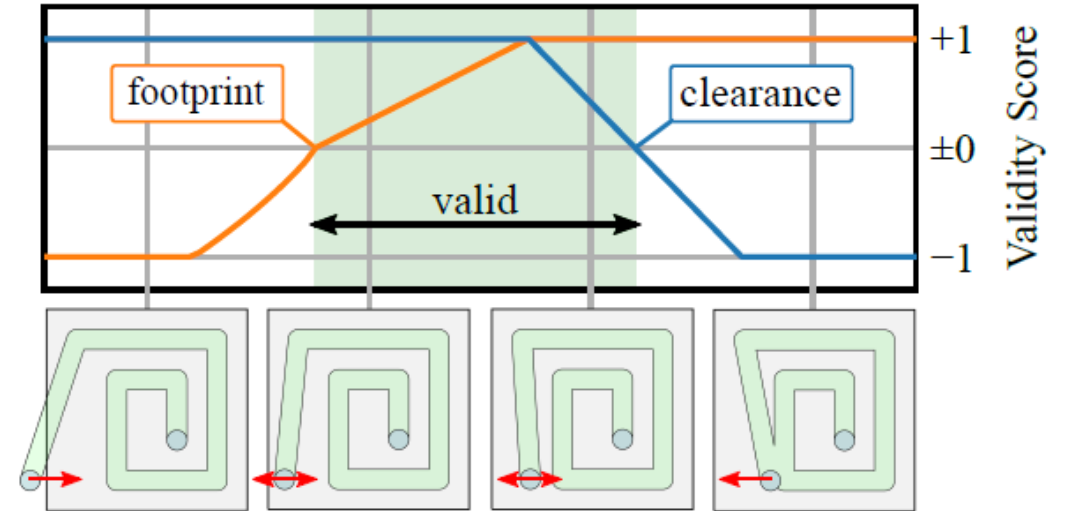
Curvature



Clearance

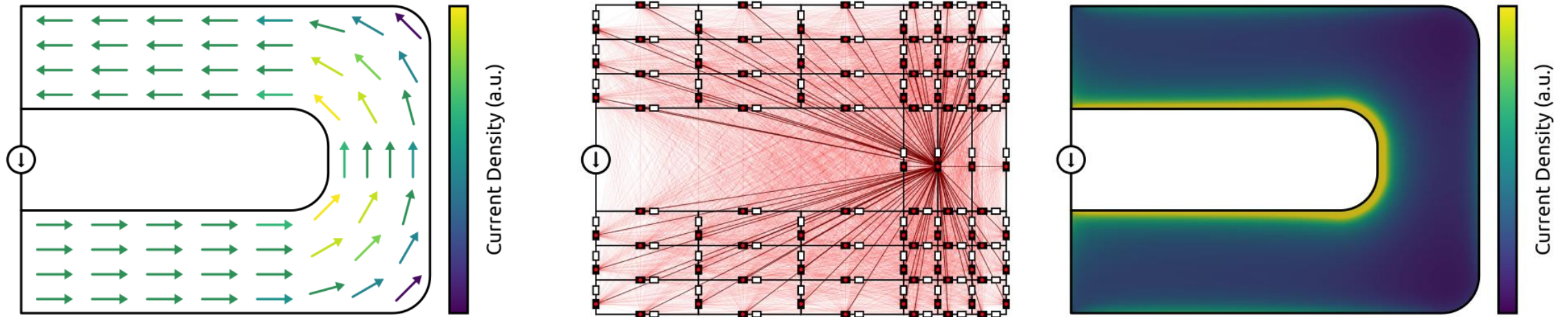


Continuous Constraints



# 3D PEEC Solver

- **PEEC (Partial Element Equivalent Circuit) method [1970s]**
  - Integral equations method using large equivalent circuit
  - Fast for air-core structures (only the conductors are meshed)
- **Generates large dense matrices**

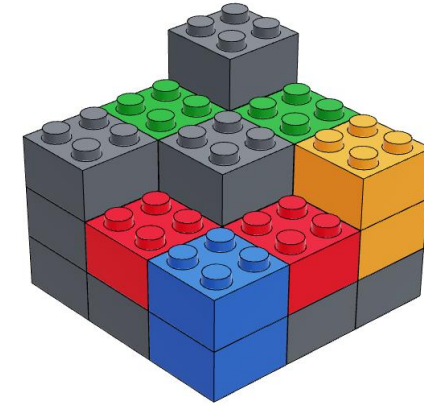




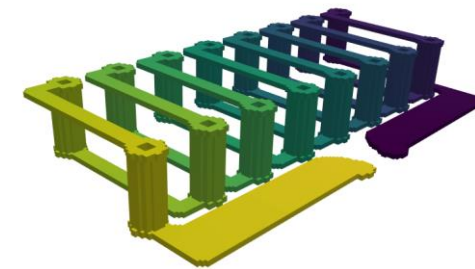
# 3D PEEC Solver

- **FFT acceleration [2018, 2022]**
  - Using **voxels** to represent the geometries
  - Embedding the matrices in **circulant tensors**
- **Advantages**
  - **Memory** storage: from  $O(n^2)$  to  $O(n)$
  - Matrix **multiplication**: from  $O(n^2)$  to  $O(n \ln(n))$
- **PyPEEC** (custom implementation)
  - **3D magnetic solver** (DC and AC)
  - Can handle **arbitrary geometries**
- **Can solve 6.5 designs per seconds**

Voxel Structure

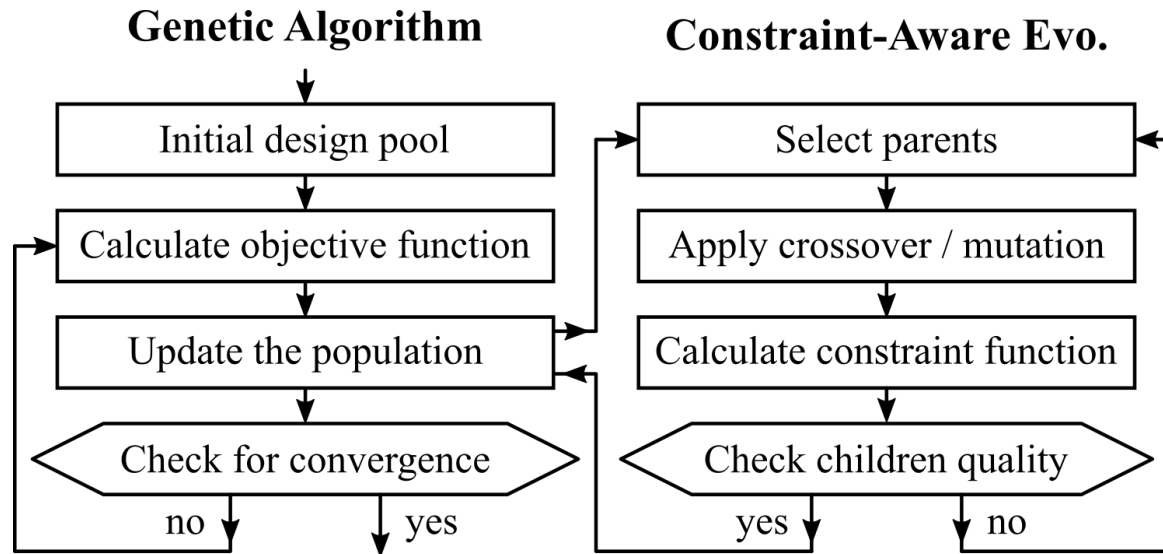


PyPEEC Inductor



# Global Optimization

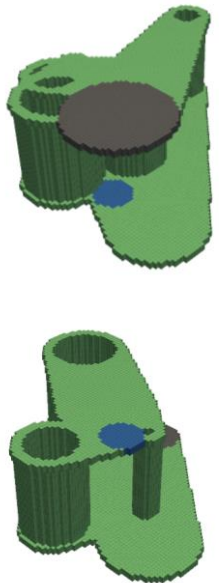
- **Gradient-free algorithms:** differential evolution, particle swarm, Parzen estimator, CMA-ES, NglohTuned, and **genetic algorithm**
- **The design rules** are extremely **restrictive**
- **Constraint-aware genetic algorithm**
- **Enforce** the design **rules** during the optimization



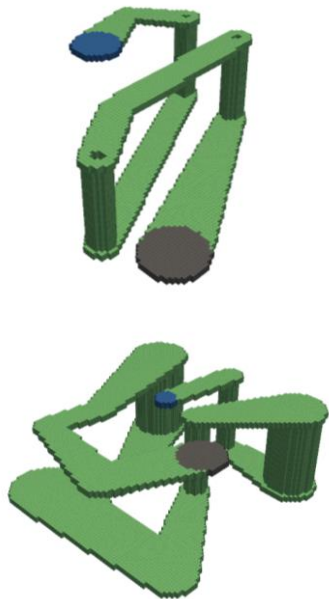
# Design Rule Awareness

- **Generation of random designs for the initial population**
  - **Probability** to obtain a **valid** design: below **one in a billion**
  - **Fully random generation is not feasible**
- **A recursive tree algorithm has been developed**

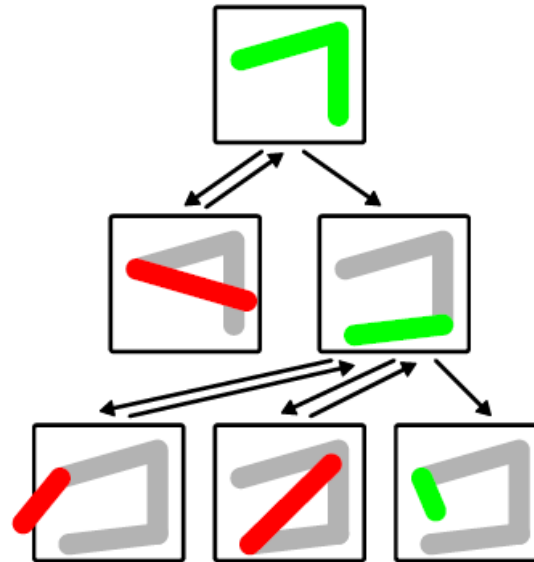
Invalid Designs



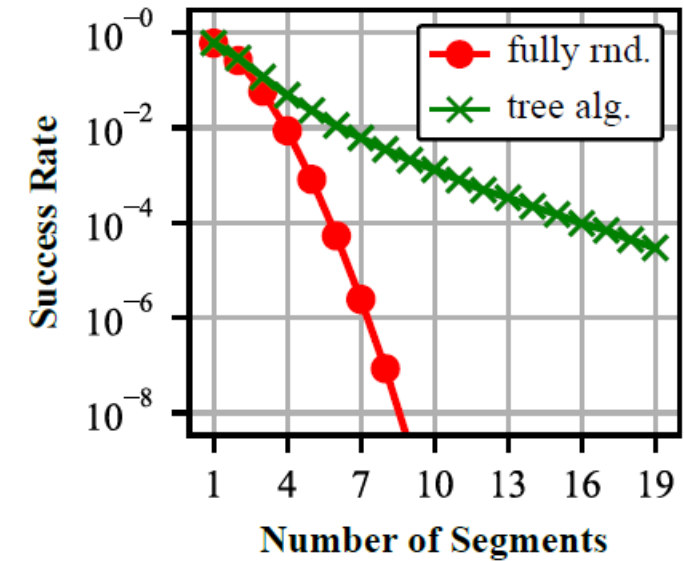
Valid Designs



Recursive Tree Example



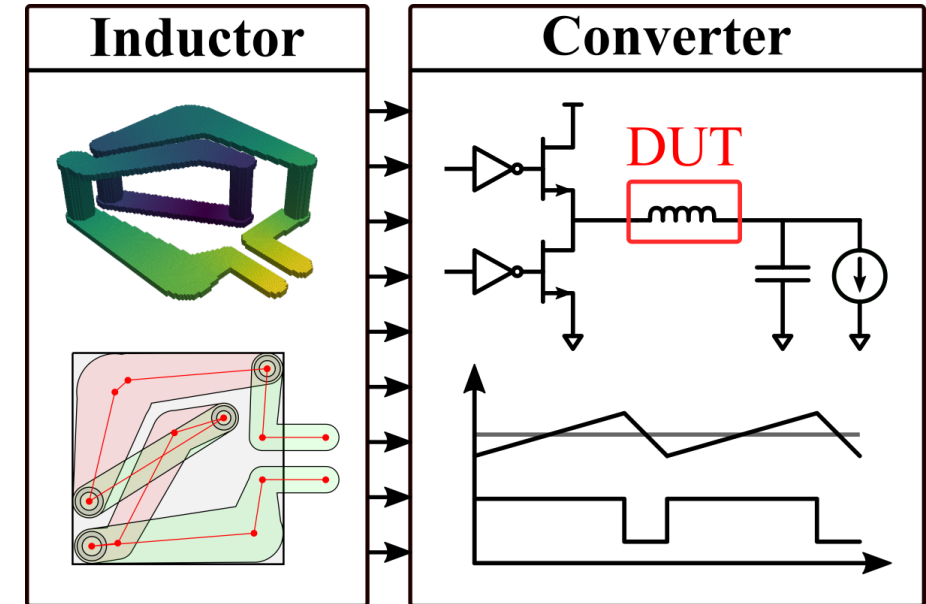
Valid Design Generation



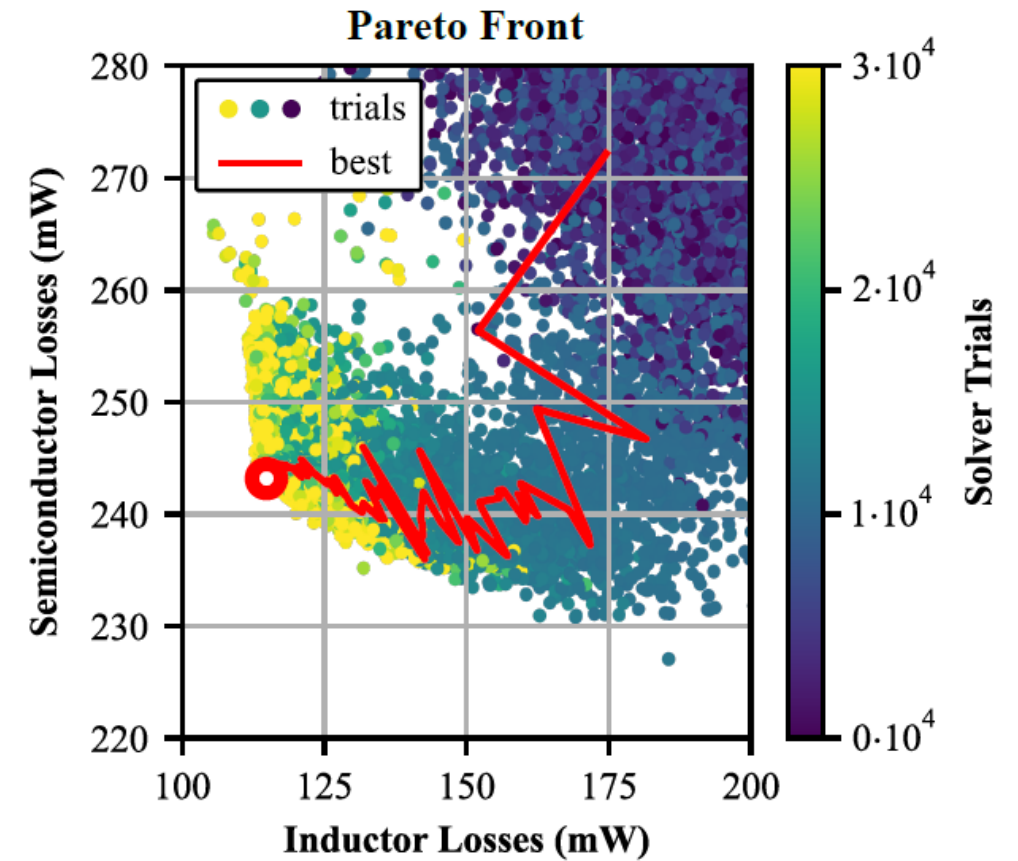
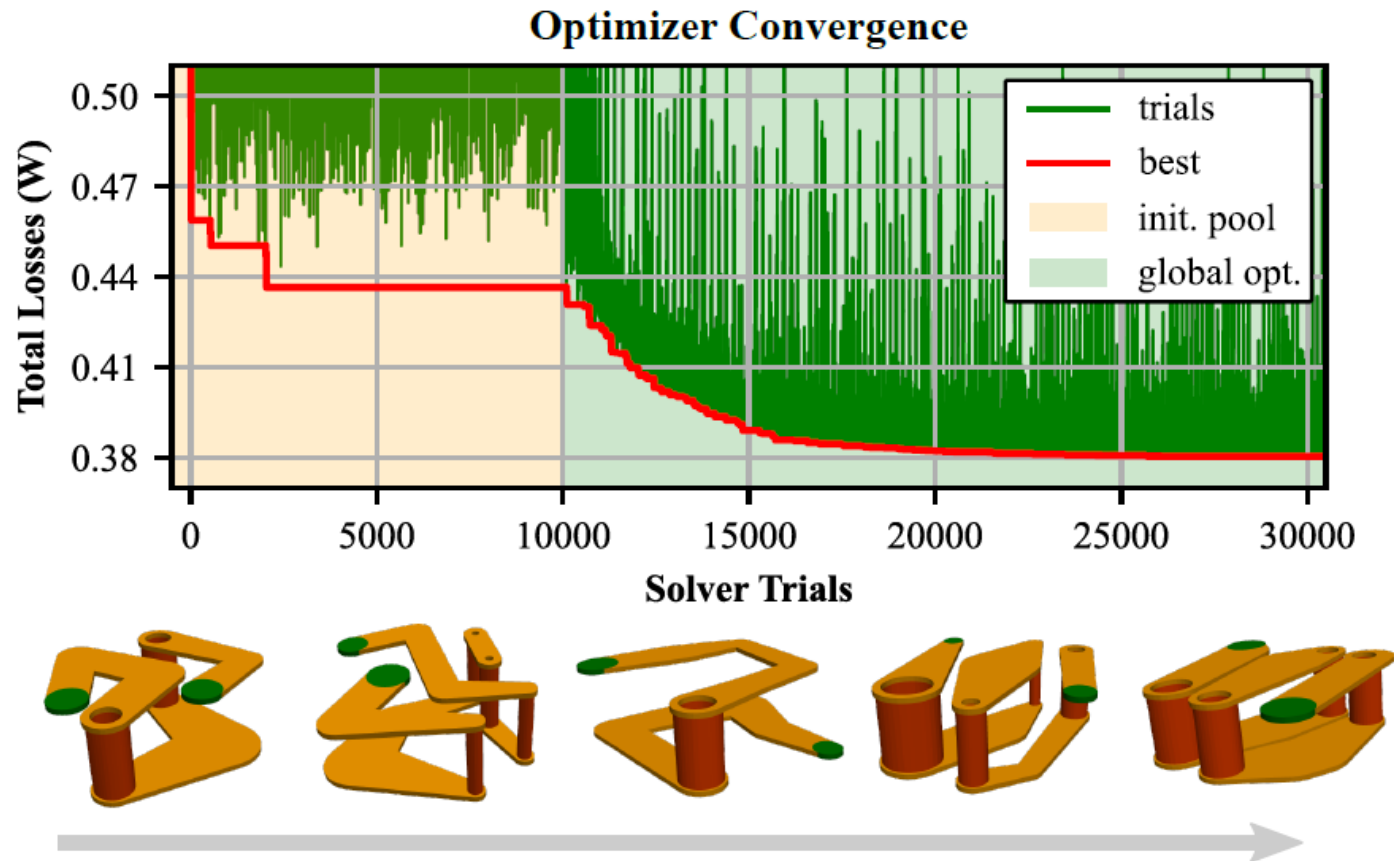
# Optimization Results

# Problem Specifications

- **Inductor of a DC-DC Buck IVR**
  - 3.3 V to 0.8 V at 1.6 W
  - 40.68 MHz operating frequency
  - 180 nm SOI switches
- **Design rules**
  - Limited footprint: 1 mm<sup>2</sup>
  - **Two-layer** planar inductor
- Minimize **converter losses**



# Optimizer Convergence



# Design Space Diversity

- **Standard geometries**

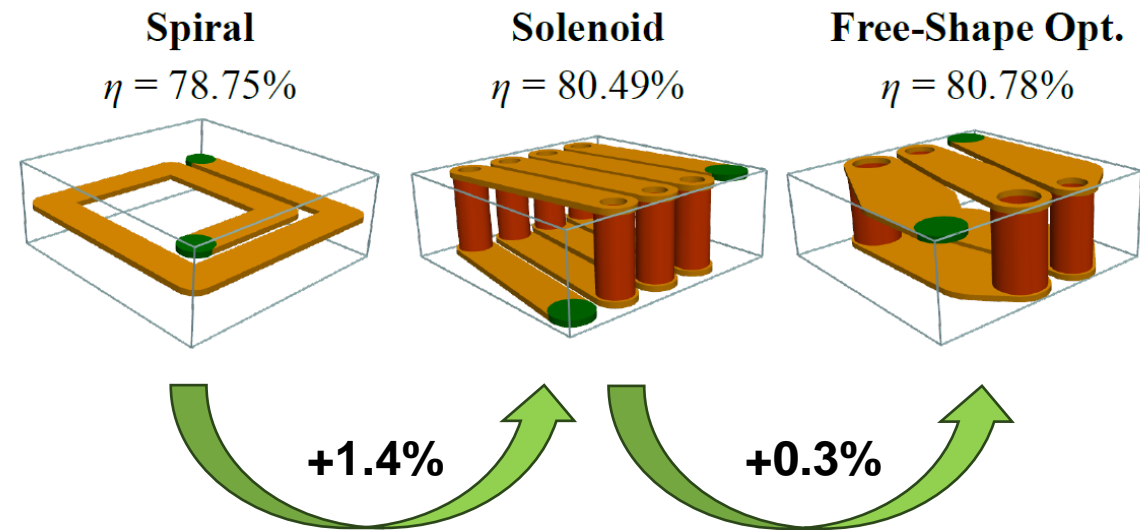
- **Spiral** with 1.5 turns
- **Solenoid** with 4 turns

- **Free-shape design**

- **Solenoid** with 3 turns
- **Only 0.3% more efficient**

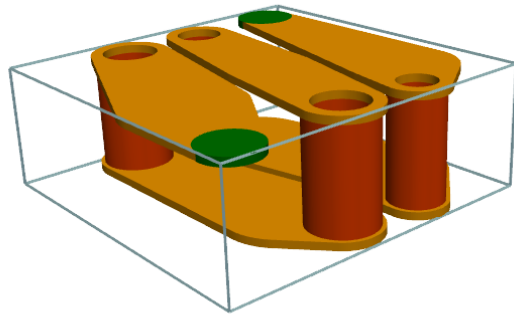
- **Is shape opt. useless?**

- Design space diversity
- Additional constraints

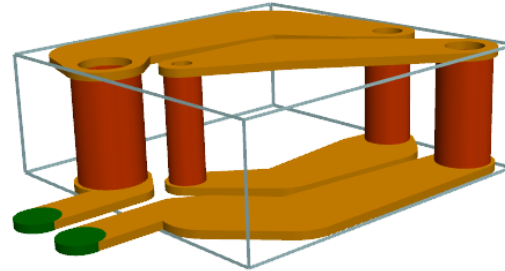


# Additional Constraints

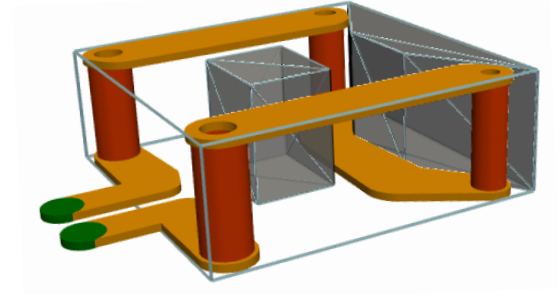
**Floating Terminals**



**Fixed Terminals**



**Footprint Constraint**



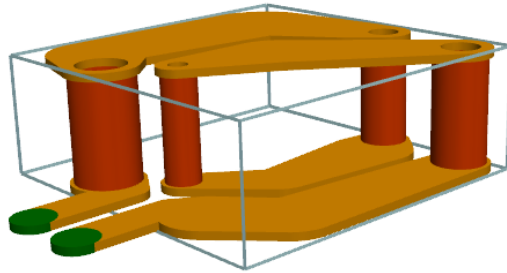
Shape	Solenoid shaped		Solenoid shaped		Loop shaped	
Total Efficiency	80.8%	- 1.8%	79.0%	- 1.2%	77.8%	
Inductor Efficiency	93.3%		90.8%		89.7%	

- **Shape optimization can handle complex constraints**
- **Terminals have a non-negligible impact**

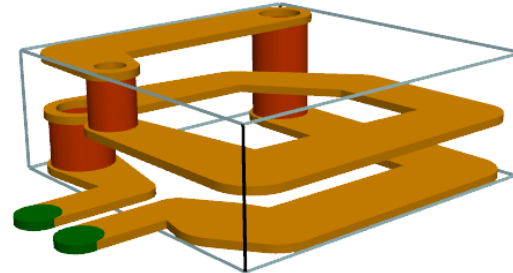


# Extension of the Design Space

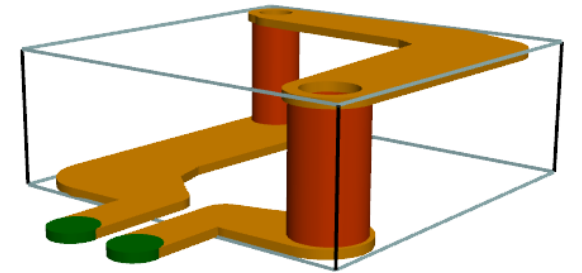
Reference Geometry



Three-Layer Geometry



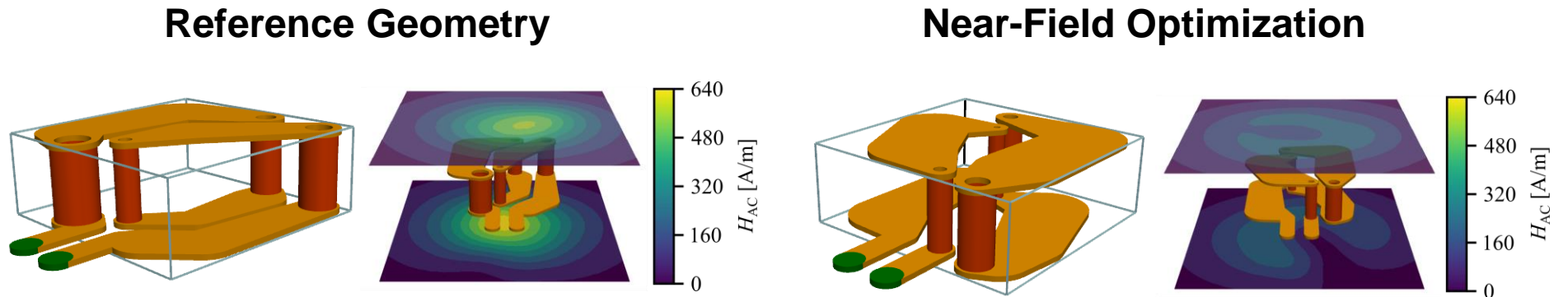
Three-Level Topology



Shape	Solenoid shaped		Loop shaped		Loop shaped	
Total Efficiency	79.0%	+ 1.0%	80.0%	+ 3.8%	83.8%	
Resistance Value	16.9 m $\Omega$		21.8 m $\Omega$		11.1 m $\Omega$	
Inductance Value	2.17 nH		3.42 nH		1.31 nH	

- Different boundary conditions lead to different shapes

# Near-Field Optimization



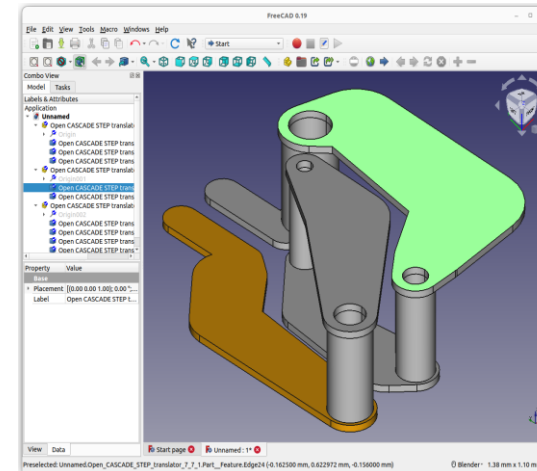
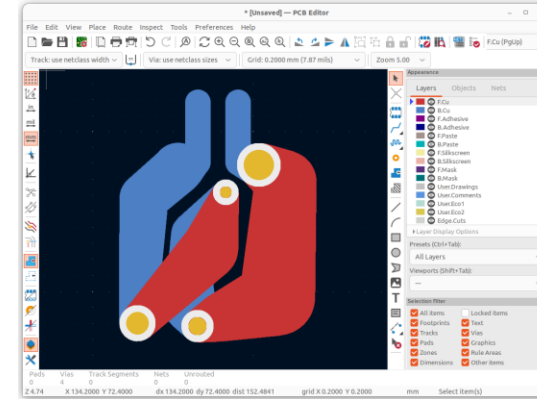
Shape	Solenoid shaped		Non-conventional shape
Total Efficiency	79.0%	- 0.9%	78.1%
DC Mag. Near Field	771 A/m		288 A/m
AC Mag. Near Field	616 A/m	÷ 2.5x	249 A/m

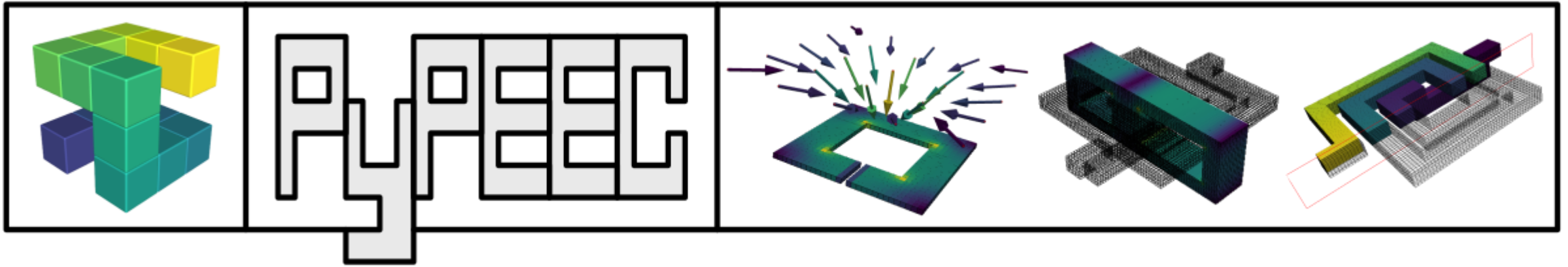
- **Magnetic near-field: eddy-current losses and/or EMI issues**
- **Massive field reduction with marginal impact on the efficiency**

# Conclusion

# Conclusion and Outlooks

- **Free-shape optimization**
  - Strict **design rule** enforcement
  - Fast **3D** FFT-accelerated **PEEC**
  - **Constraint-aware genetic algorithm**
- **Air-core inductors for IVRs**
  - Classical shapes good for standard problems
  - Shape optimization can handle complex constraints
- **Outlooks**
  - Scaling to more complex problems
  - Gradient-based methods (e.g., auto-diff)
  - Neural networks (e.g., surrogate)

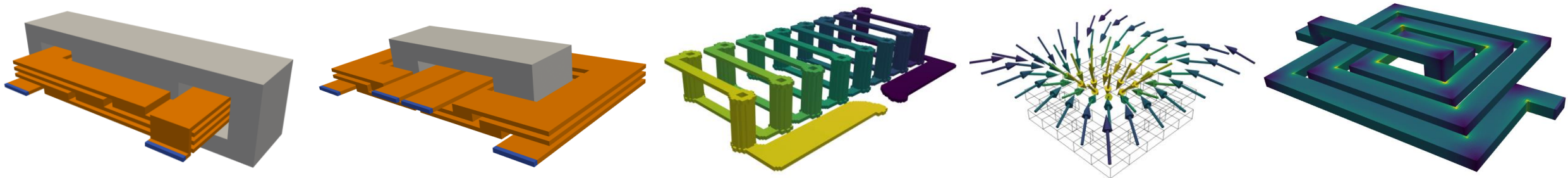




- 3D FFT-accelerated PEEC solver
- DC and AC magnetic problems
- Fully open source Python code



[pypeec.otvam.ch](https://pypeec.otvam.ch)



# Shape Opt. Source Code



[github.com/otvam/pyfreecoil](https://github.com/otvam/pyfreecoil)



[zenodo.org/records/14247697](https://zenodo.org/records/14247697)



# APEC 2025



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## Thank you! Questions?



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