

# Ready for AI?

Building surrogate models via  
Cloud-native AI simulation

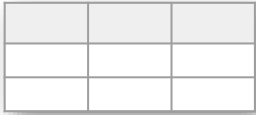
Speaker



**Alexander Fischer**  
PM & Co-founder -  
SimScale

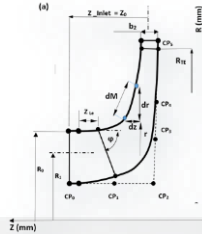


# Many steps in the engineering design process will be AI-augmented



## Requirements

- $Q$  = Flow rate ( $\text{m}^3/\text{h}$ )
- $H$  = Head (m)
- $n$  = Rotating speed (rpm)
- $P$  = Motor power (kW)
- $\eta = Q \cdot H / P$  = Efficiency (%)
- ...



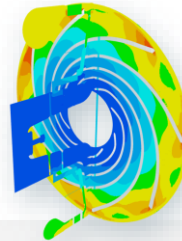
## Concept 1-2D Design

- Rough hydraulic design
- Analytical calculations via Euler equations, no losses included
- ...



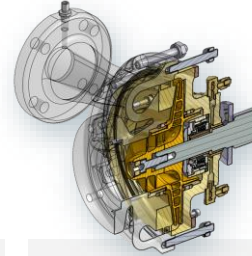
## Concept 3D Design

- Just hydraulic parts, no gaps
- Main CAD parameters:
  - $\beta$  = inlet angle
  - $\alpha$  = outlet angle
  - Number of blades
  - ...



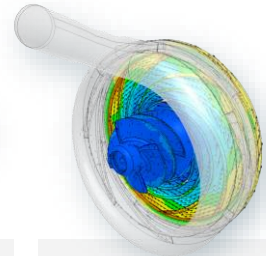
## Early 3D Simulation

- Early feedback on hydraulic performance
- Early feedback on structural performance
- Inform concept design



## Detailed Design

- "As built"
- All gaps / details included



## Final 3D Validation

- Final hydraulic performance validation
- Final structural performance validation

## AI Use Cases (Examples)

### Requirements Engineering

LLMs help structure requirements faster and cleaner and break it down into atomic requirements.

Example: [Valispace AI](#)

### Simulation Automation

Procedures during the simulation setup will be automatically completed / suggested based on previous human completion.

Example: [ADSK Assembly Mating](#)

### AI Simulation

AI models trained on design-specific, small amounts of simulation data can be used to replace PDE-based solvers.

Example: [Navasto FEA](#)

### Design Automation

Based on requirements, geometry can (partially) be generated automatically using LLMs and CAD APIs, implicitly building a true digital thread

Example: [BlenderGPT](#)

### Instant Result Preview

AI models trained large amounts of generic simulation data can be used to preview PDE-based solver results.

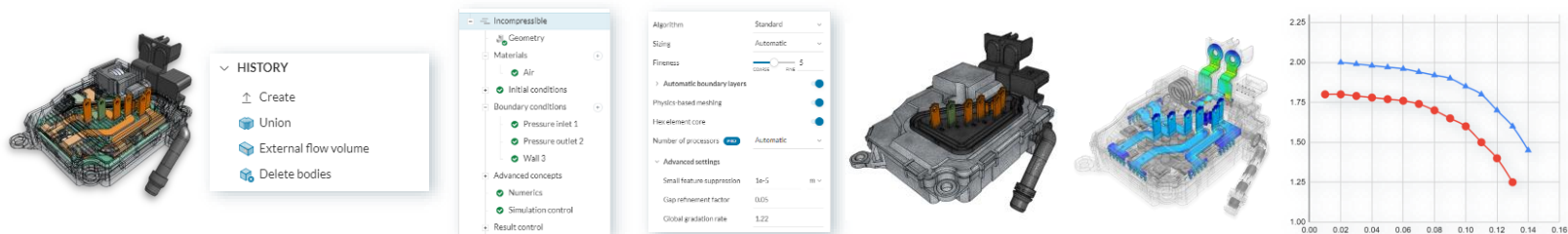
(Currently in development at SimScale)

### AI Optimization

Surrogate models derived from simulation data can be used to optimize without geometry creation at all.

Example: [KSB](#), [Caeses](#) & [SimScale](#)

# SimScale's architecture provides a mature basis for AI training and inference



Raw Data

CAD

CAD Edit

Simulation Setup

Mesh Setup

Mesh

3D Results

Scalar Results

Aggregation

*Data Management in support of training and inference. GPUs for training. AI Model lifecycle management.*

AI Support

Roadmap

**Pre-Proc Autom.**

- Automatic CAD cleaning
- Automatic meshing

In development

**Setup Suggestion**

- Suggested setup
- Best template to use

In production

**Resource Prediction**

- Compute Runtime
- Compute resources

In production

**Results Preview**

- 3D Results
- Scalar Results

In production

**AI Simulation**

- 3D Results
- Scalar Results

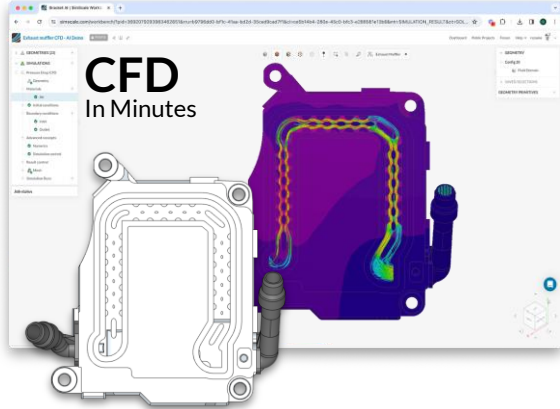
Roadmap

**Design Insights**

- Problem areas
- Alternative design gen

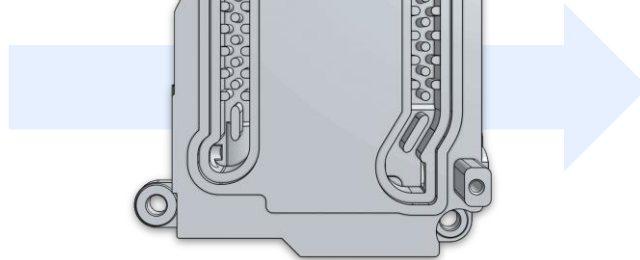
Reducing simulation lead time

# From **hours** to **seconds**: Enable accurate real-time simulation via AI



## Cloud-Native CFD

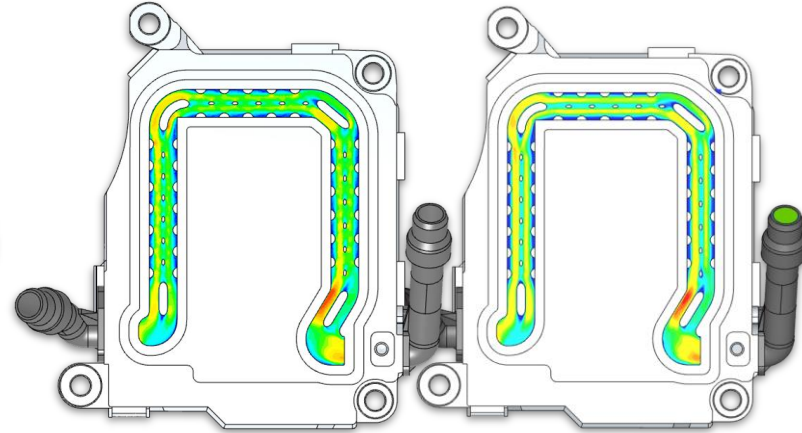
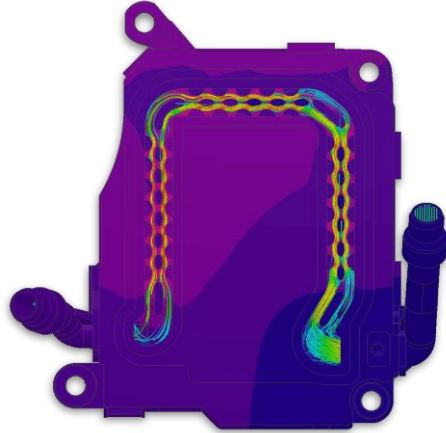
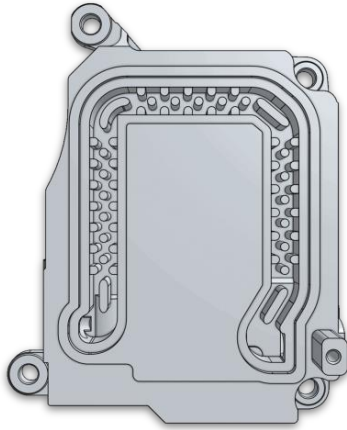
- All simulations immediately ready for AI training
- AI training on cloud GPUs
- As new data is added, re-training is straightforward



## Cloud-Native AI CAE

- AI CAE Models are deployed right next to CFD simulations
- AI used for fast evaluations, CFD available for validation

# AI Preview Demo: EV Power Inverter



CFD Reference

AI Prediction

## EV Power Inverter Model

- MOSFET temperature & pressure drop
- Geometry variations for turbulence elements in:
  - Shape (8 variations)
  - Thickness (10+ variations)

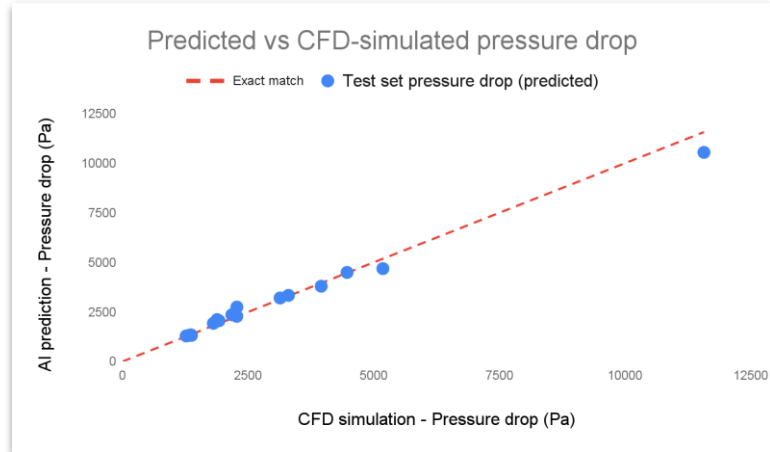
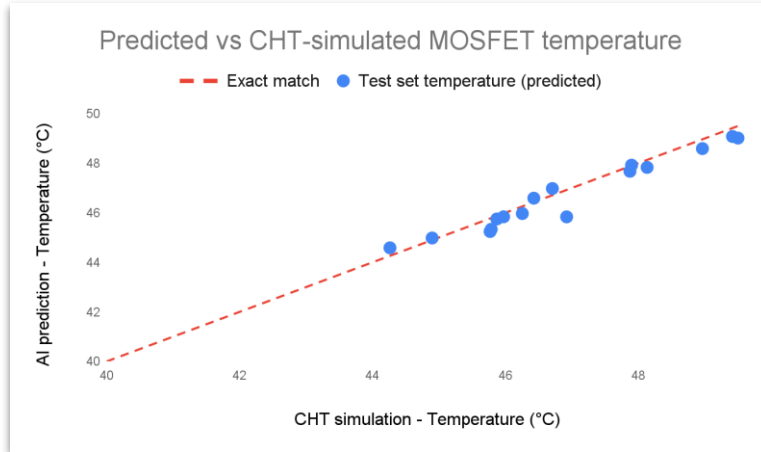
## CFD via API on SimScale

- Run all 104 variants in parallel on 16 core instances
- Complete dataset generation within 25 minutes, 15 min avg

## AI model generation and deploy

- Model generation with fully data-driven Graph Neural Network (GNN)
- Automatic hyperparameter optimization, best model selection
- Total runtime <24h on single GPU

# AI Preview Demo: EV Power Inverter



## AI Prediction test set

- 89 of 104 simulations used to **train** the AI model
- 15 configurations used as **test set**
- Comparison of AI prediction and simulation result on 'unseen' test set

## AI Prediction confidence

- Test set **accuracy** for results of main interest
- **Confidence** interval describing configuration fit to training data
- Additional **metrics** indicating max and avg errors in test set configurations



**Thank you for your  
time!**

