

Aerodynamic optimization of low pressure axial fans with OpenFOAM

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Optimization of low pressure axial fans

Outline

- 1 Introduction
- 2 Optimization with Optimus[®] and OpenFOAM
- 3 Some results
- 4 Summary and outlook

1 Introduction

Objective and Motivation

- Axial fans are typically used to pump air through the car underhood
- Cooling fans belong to the highest electrical consumers in conventional cars (up to ~1200 W)
- While the needed vol. flowrate is quite high ($> 2500 \text{ m}^3/\text{h}$), the needed pressure rise is rather low ($< 300 \text{ Pa}$)

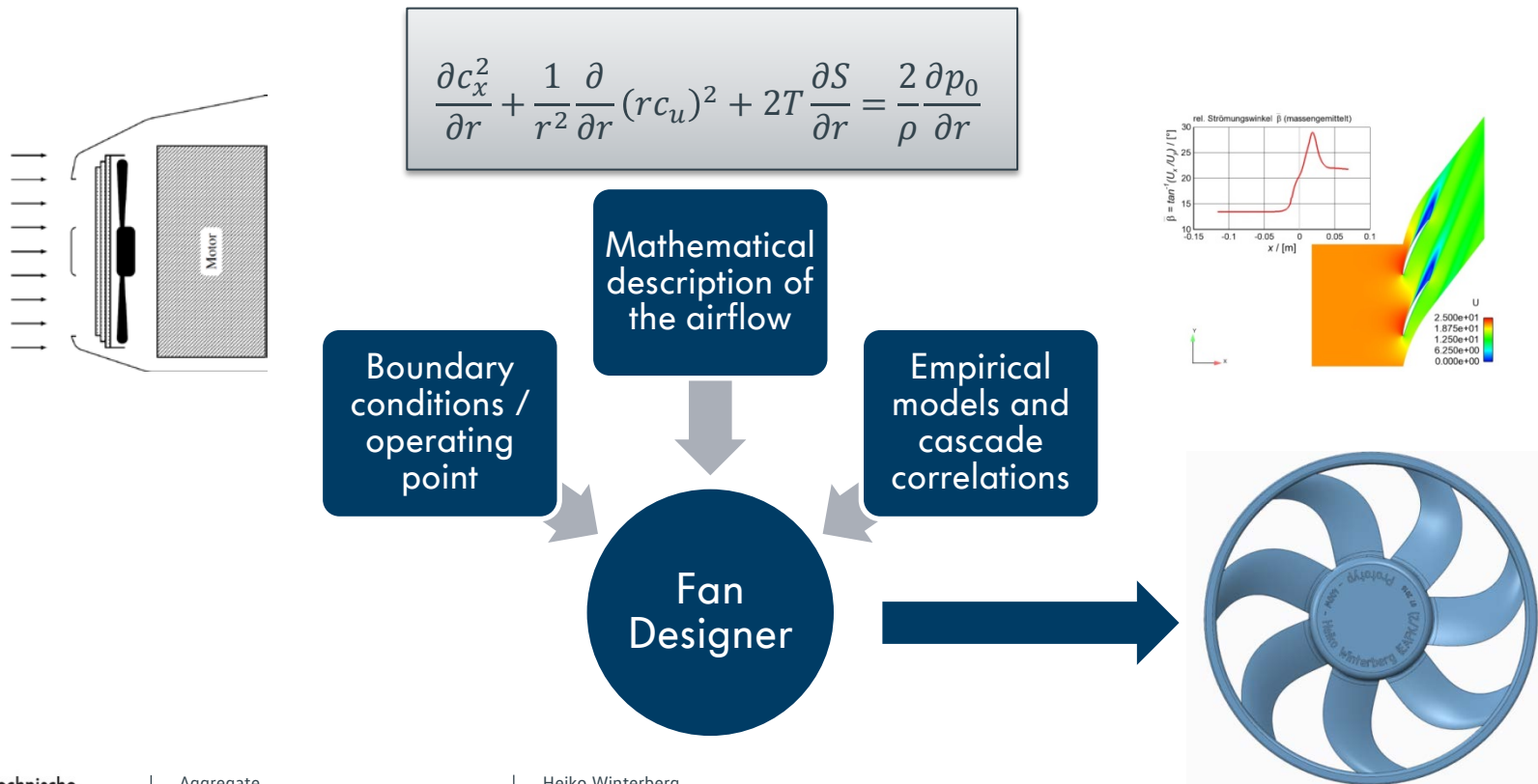
Aim

- Lower max. power consumption or
- Increase vol. flowrates at specific operating points

1 Introduction

Design Principle of Axial Fans

Development of a MATLAB™-program for axial fan design



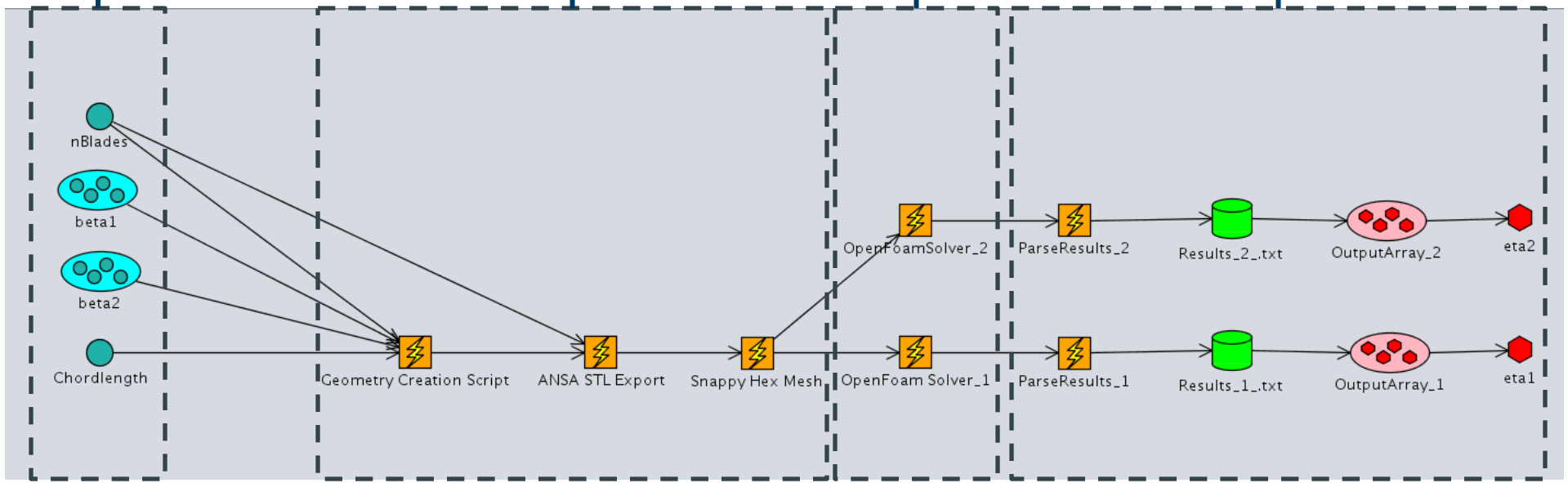
2 Optimization Overview

Parameterset
Optimus / Matlab

Preprocessing
Matlab / ANSA / OpenFOAM

3D-CFD
OpenFOAM

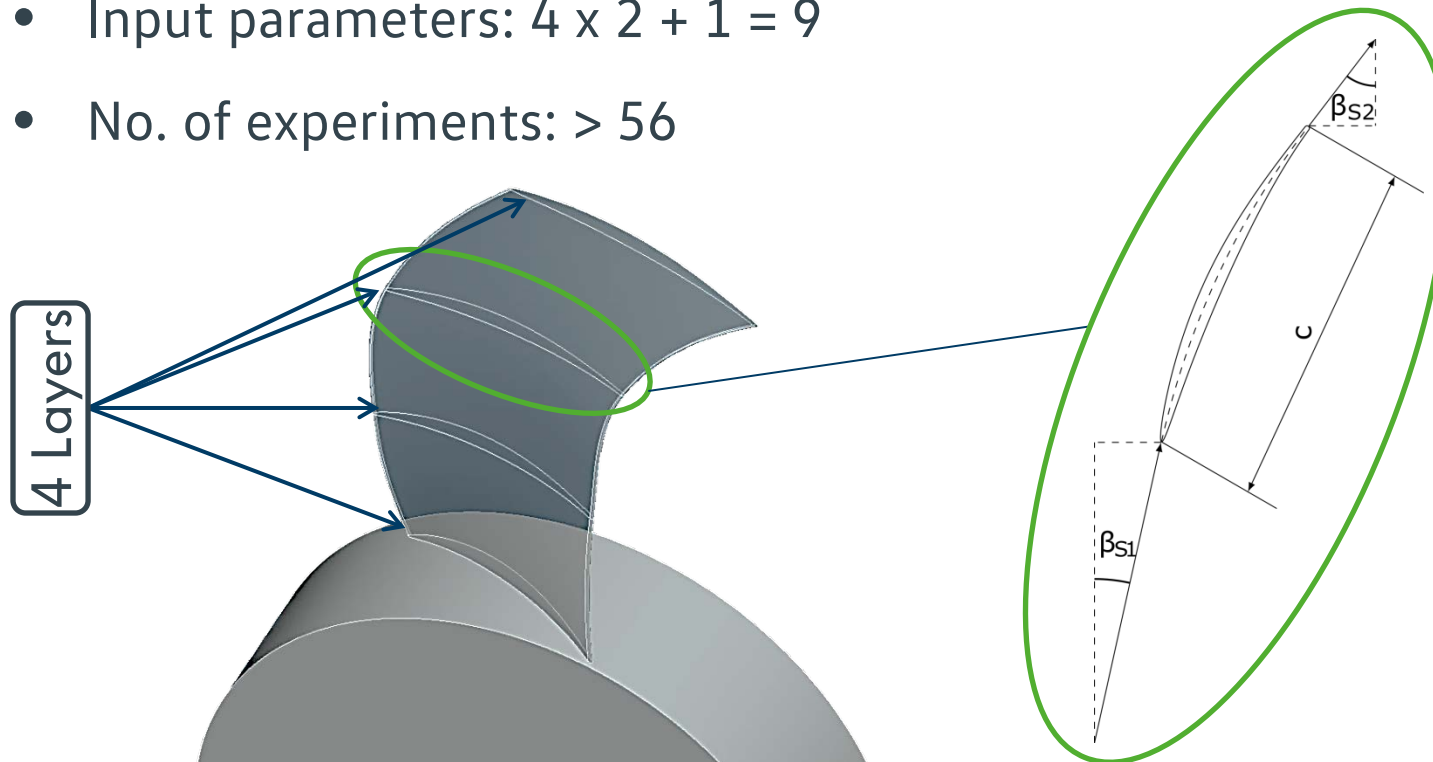
Postprocessing
OpenFOAM / (EnSight) / Optimus



2 Optimization

Design of Experiments

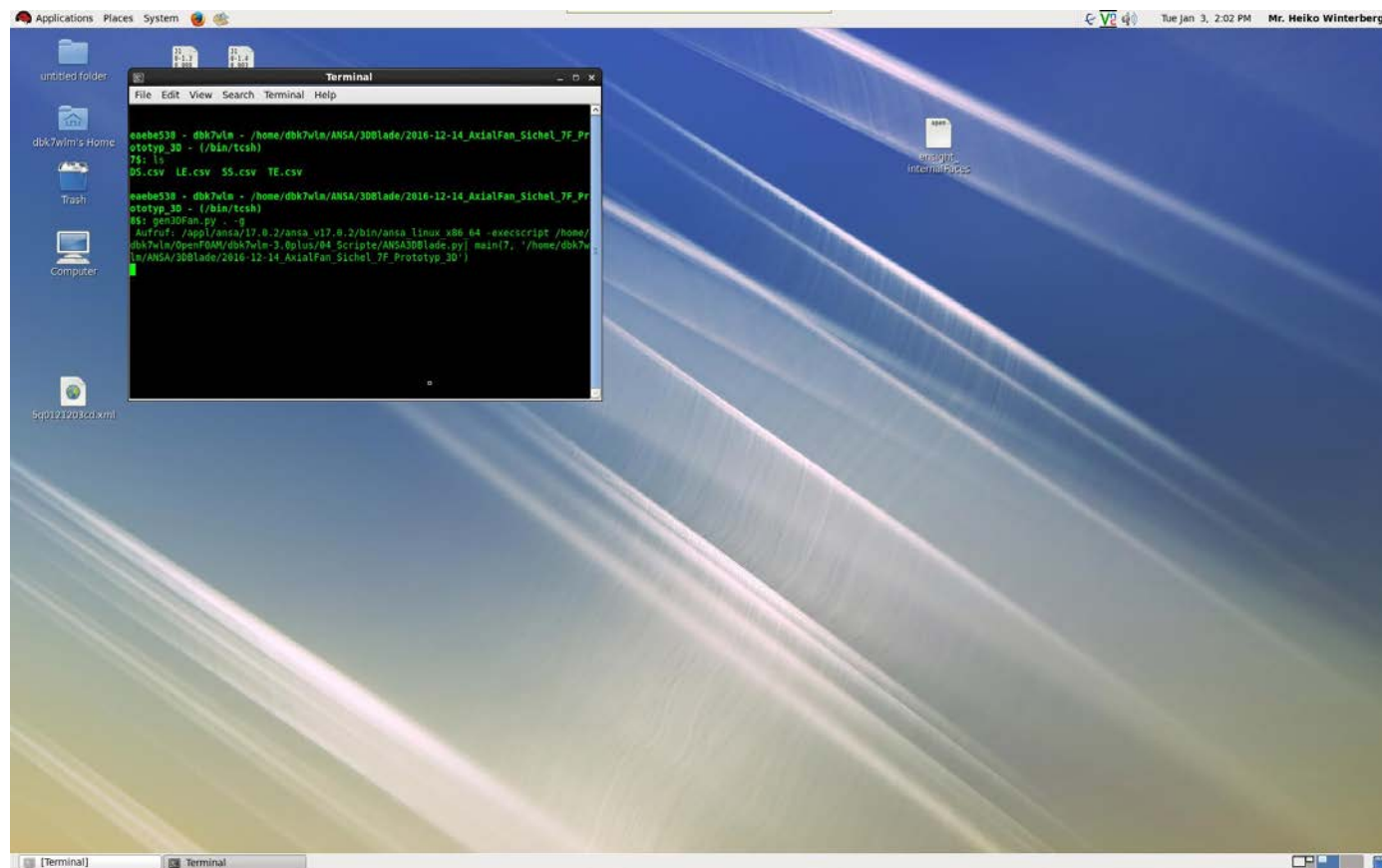
- Optimus[®] used to generate the DoE (Latin Hypercube)
- Input parameters: $4 \times 2 + 1 = 9$
- No. of experiments: > 56



2 Optimization

Preprocessing (3D-CAD generation)

Development of an interface between Fan Designer and ANSA™ using the Python API



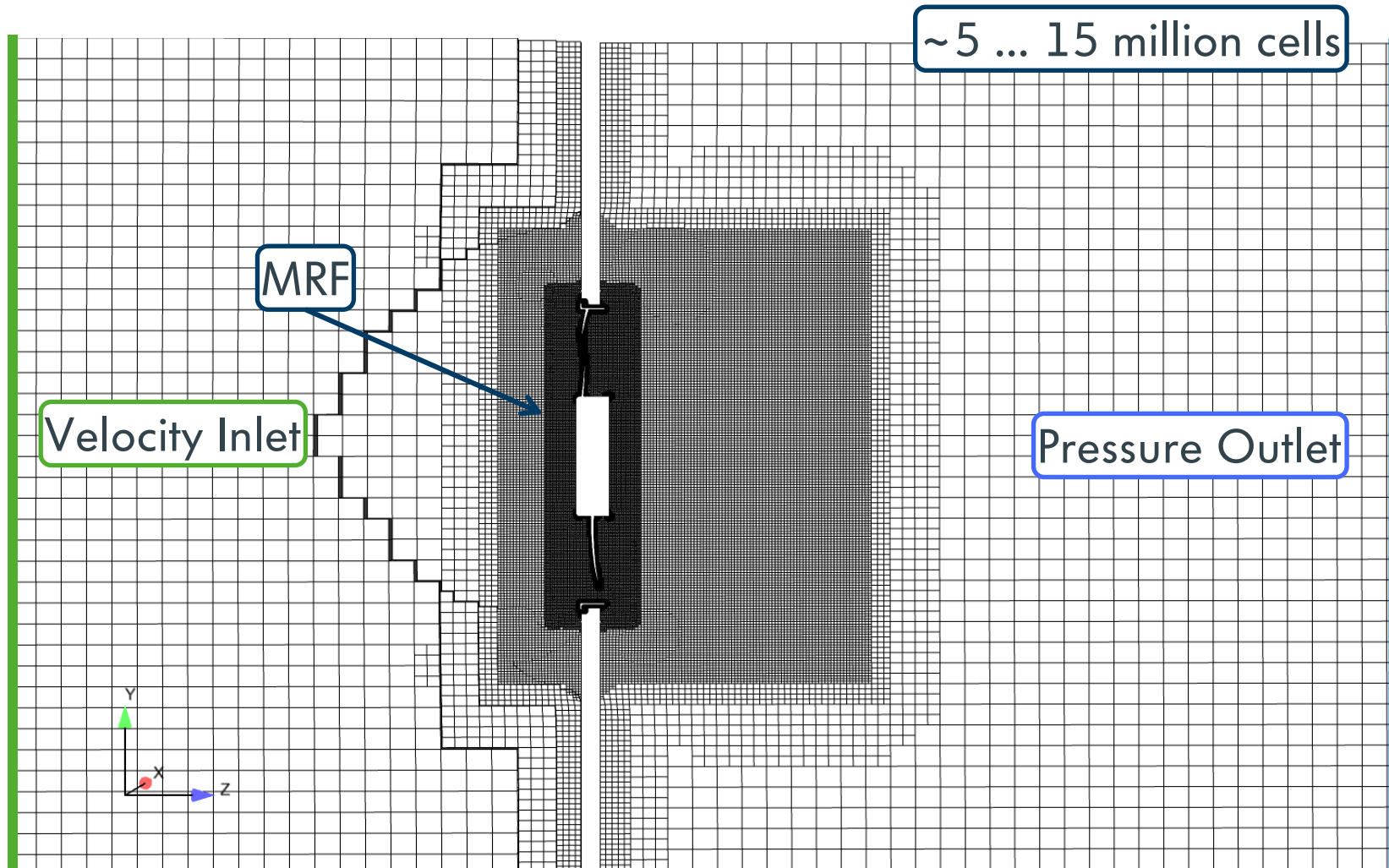
The screenshot shows a Linux desktop with a terminal window open. The terminal displays the following commands and output:

```
eaebes38 - dbk7vln - /home/dbk7vln/ANSA/3D0lade/2016-12-14_AxialFan_Sichel_7F_Pr
ototyp_3D - (/bin/tcsh)
75: ls
05.csv LE.csv SS.csv TE.csv

eaebes38 - dbk7vln - /home/dbk7vln/ANSA/3D0lade/2016-12-14_AxialFan_Sichel_7F_Pr
ototyp_3D - (/bin/tcsh)
05: gen3DFan.py - -g
  xitrcsf: /app1/ansa17.0.2/ansa.v17.0.2/bin/ansa linux x86_64 -execscript /home/
dbk7vln/OpenFOAM/dbk7vln-3.0plus/04_Scripts/ANSA3D0lade.py main(7, /home/dbk7v
ln/ANSA/3D0lade/2016-12-14_AxialFan_Sichel_7F_Prototyp_3D*)
```

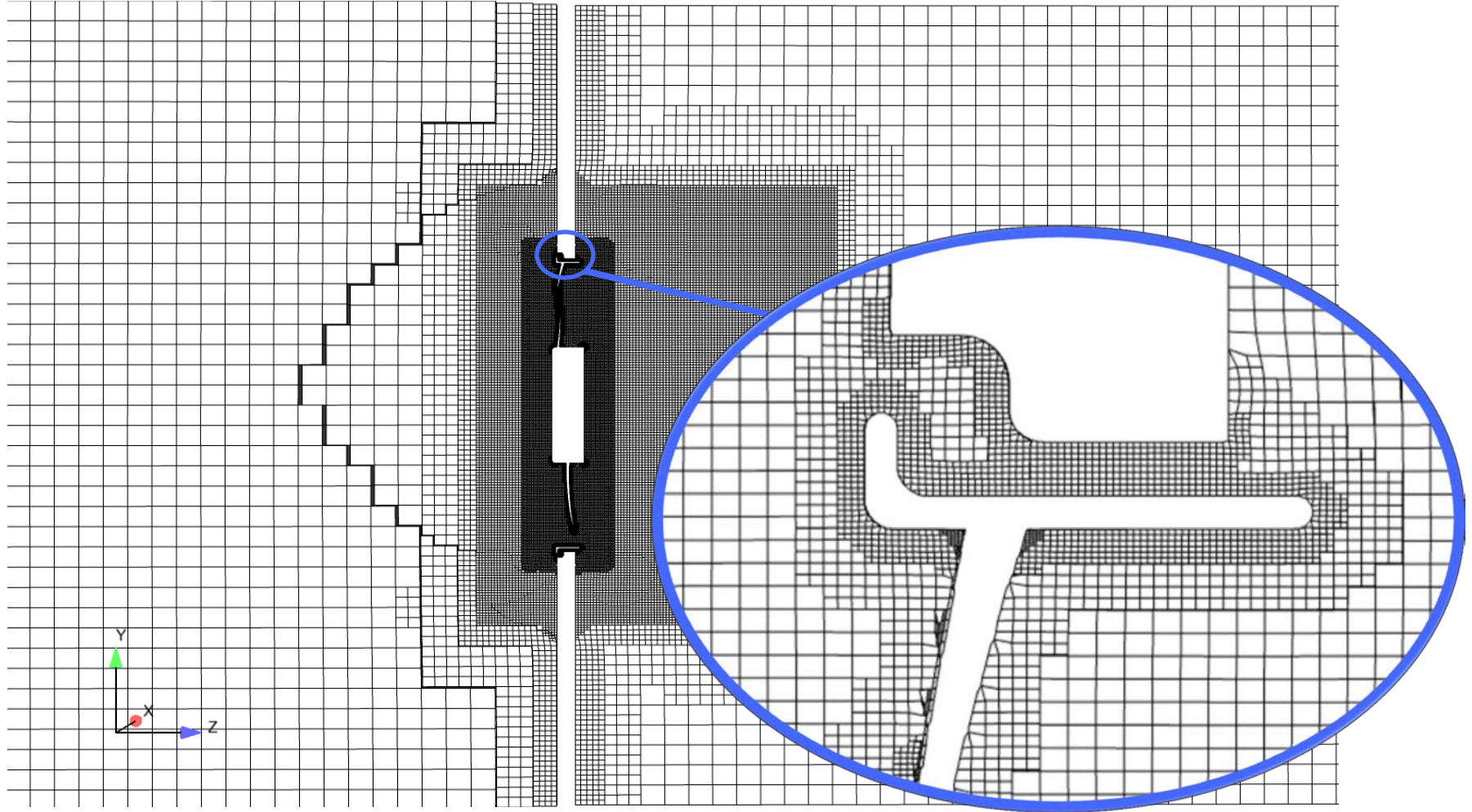

2 Optimization

Preprocessing (Meshing with shm)



2 Optimization

Preprocessing (Meshing with shm)



2 Optimization

3D-CFD

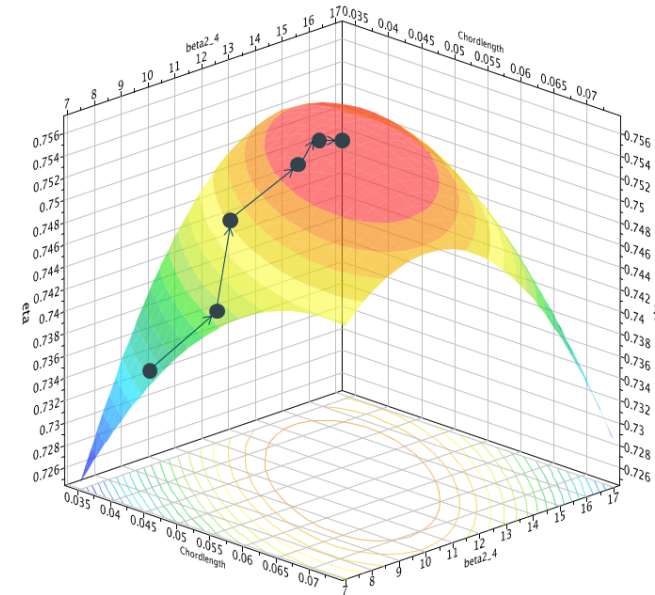
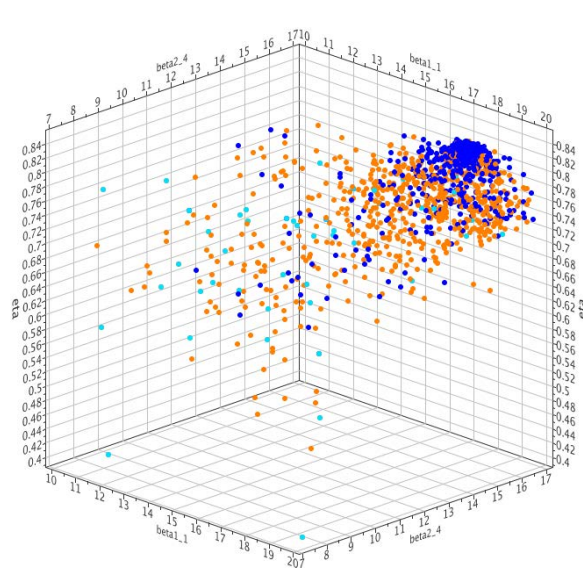
- Solver: simpleFoam
 - steady state
 - frozen rotor (MRF) approach
 - kOmegaSST (optional: kklOmega) turbulence model
 - Running parallel on 64 cores
- Automatic case setup using a python-script
- Runtime max. 10 h

2 Optimization

Postprocessing

- Fully automatic postprocessing using the OpenFOAM toolbox coupled with a python script
- Export of the integral results (\dot{V} , Δp , η , ...) into *.txt, *.PPTX and *.XLSX
- Import and storage of the results in Optimus[®]

2 Optimization Strategy



1. Parameter variation based on DoE (Latin Hypercube)
2. Response Surface Model (RSM) using the DoE-data
3. Find the global optimum of the RSM
4. Validation of the best fan-setup (3D-CFD)

2 Optimization

Response Surface Model

RSM using a Second Order Taylor Polynomial:

$$\begin{aligned}
 y &= \beta_0 + \sum_{i=1}^k \beta_i x_i + \sum_{i=1}^k \beta_{ii} x_i^2 + \sum_{i < j=2}^k \sum_{j=2}^k \beta_{ij} x_i x_j + \epsilon \\
 &= \beta_0 + \mathbf{x}' \boldsymbol{\beta} + \mathbf{x}' \mathbf{B} \mathbf{x} + \epsilon
 \end{aligned}$$

Where:

$$\mathbf{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_k \end{pmatrix}, \quad \boldsymbol{\beta} = \begin{pmatrix} \beta_1 \\ \vdots \\ \beta_k \end{pmatrix}, \quad \mathbf{B} = \begin{bmatrix} \beta_{11} & \beta_{12}/2 & \cdots & \beta_{1k}/2 \\ & \beta_{22} & \cdots & \beta_{2k}/2 \\ & & \ddots & \vdots \\ \text{sym.} & & & \beta_{kk} \end{bmatrix}$$

$\boldsymbol{\beta}$ is obtained by the method of least-squares.

2 Optimization

Optimization-Algorithm

Objective function:

$$\min_{\mathbf{x} \in \Omega} F(\mathbf{x}) = 1 - (\eta_1(\mathbf{x}), \dots, \eta_m(\mathbf{x}))'$$

Where Ω is the chosen parameter space, $\mathbf{x} = (x_1, \dots, x_k)'$ and η_m represents the efficiency for m different OPs

Constraints:

- Operating Point ($\dot{V}_{\text{op}}, \Delta p_{\text{op}} \pm \varepsilon$), or
- Power consumption ($\dot{V} \cdot \Delta p / \eta = c \pm \varepsilon$)

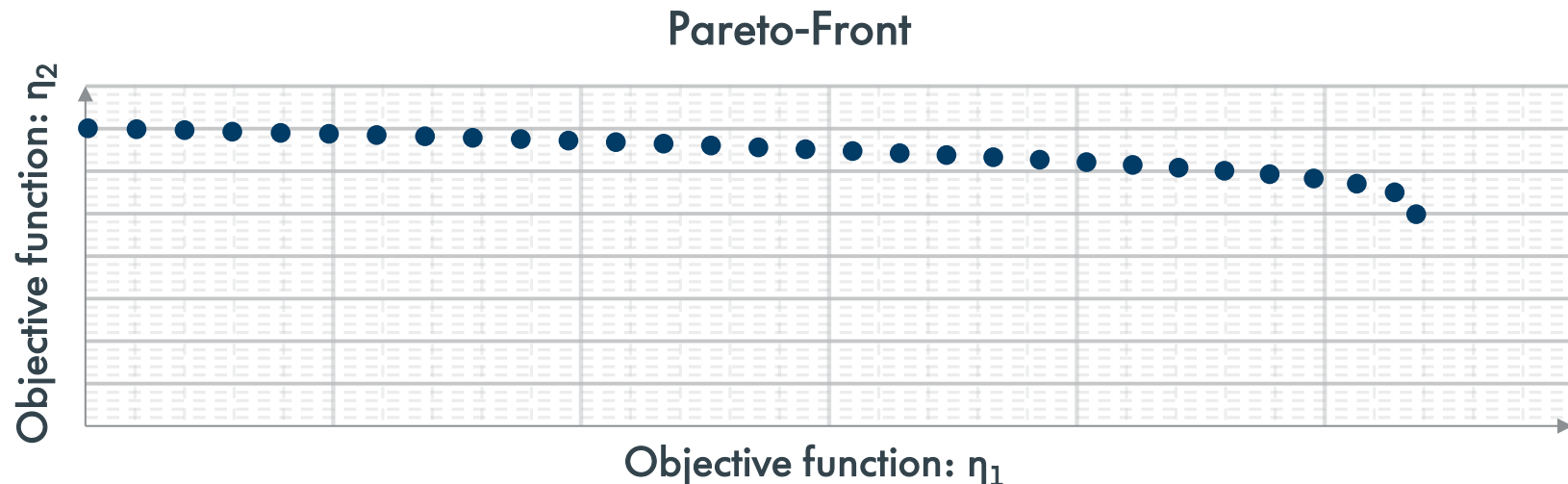
Optimization algorithm:

- Normal boundary intersection method

3 Some results

Pareto-Front

- The optimization returns the Pareto-Optimal geometries
 - Every point represents a fan-geometry, where objective function (OF) 1 can't be increased without lowering OF 2

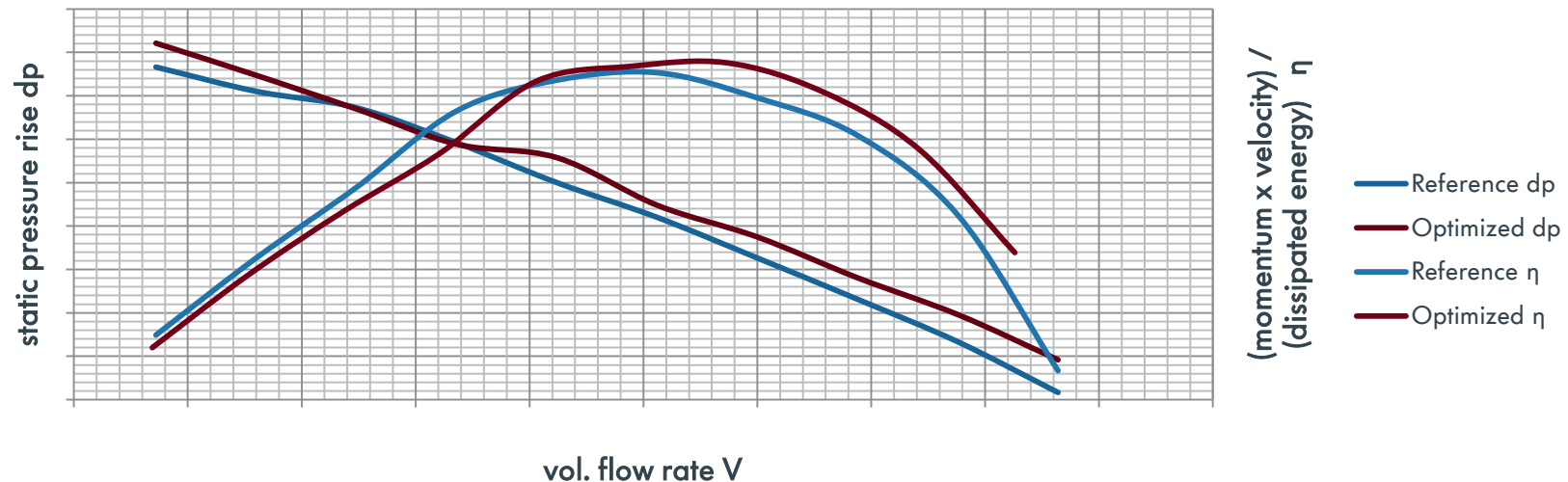


3 Some results

Fan curve

Qualitative comparison to a chosen reference fan

- Fan curve in terms of OPs optimized
- Maximum efficiency increased
- Partial load region with decreased efficiency



4 Summary and outlook

- Development of a tool for the purpose of accurate fan design
- Setup of a DoE-process using Optimus[®] coupled with Fan Designer, ANSA and OpenFOAM
- Response Surface Model of the DoE-Data using a 2nd order Taylor Polynomial
- RSM-Model based optimization to find the best design

4 Summary and outlook

Next Steps:

- Further Robust Design Analysis (fan curve optimization)
- Influence no. of experiments on RSM
- Benchmarking of different optimization algorithms
- Further validation of the process



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Backup

