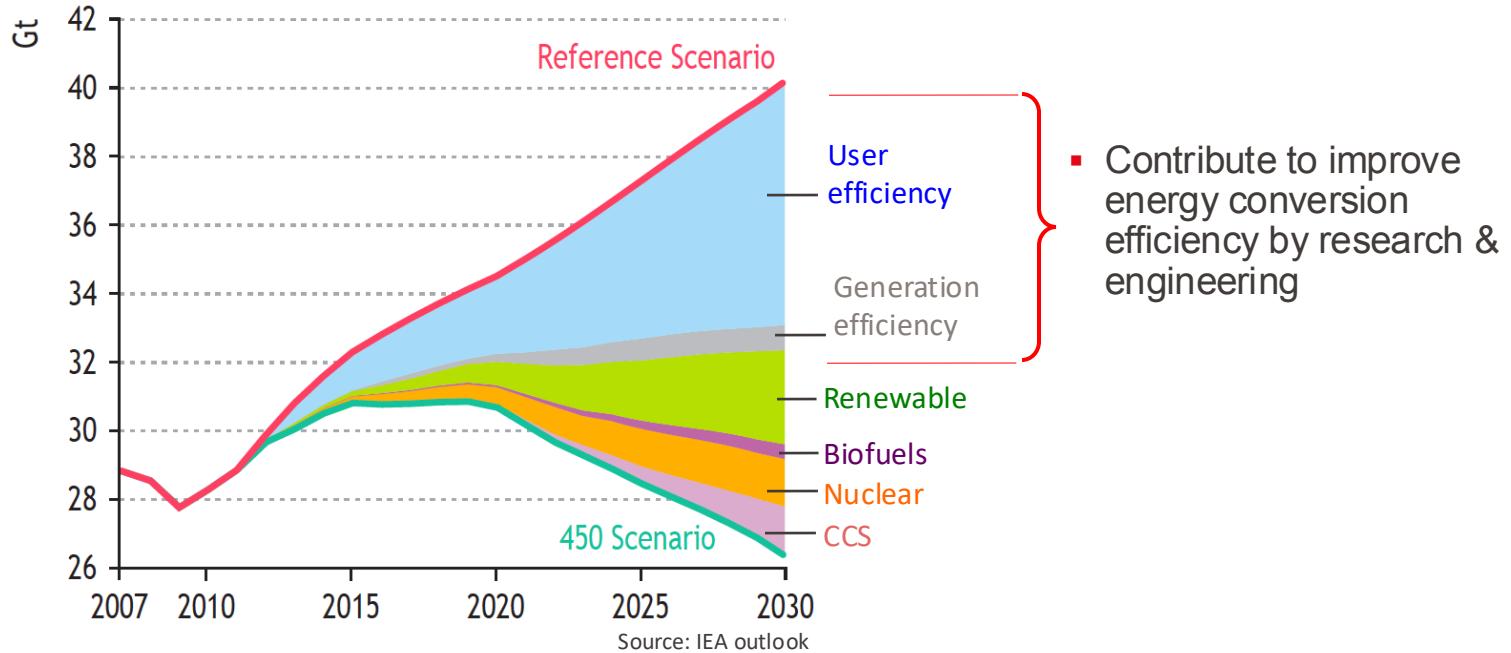


Prof. J. Schiffmann

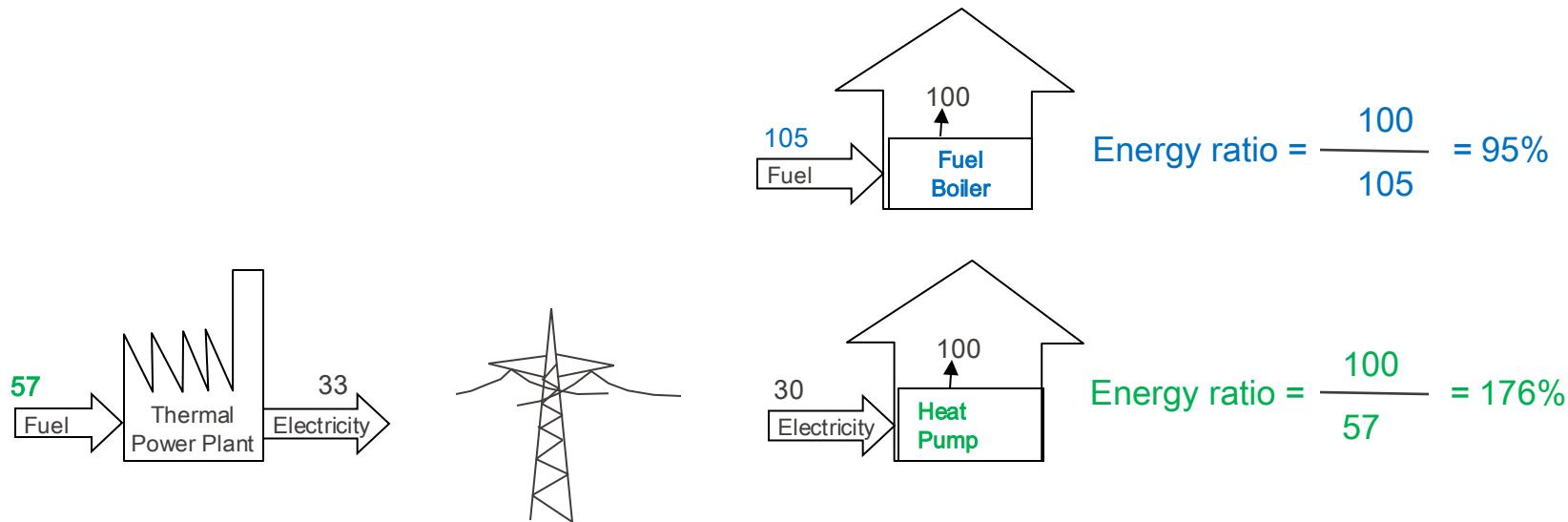
Laboratory for Applied
Mechanical Design
Department of
Mechanical Engineering

Steps Towards Sustainable Energy Landscape



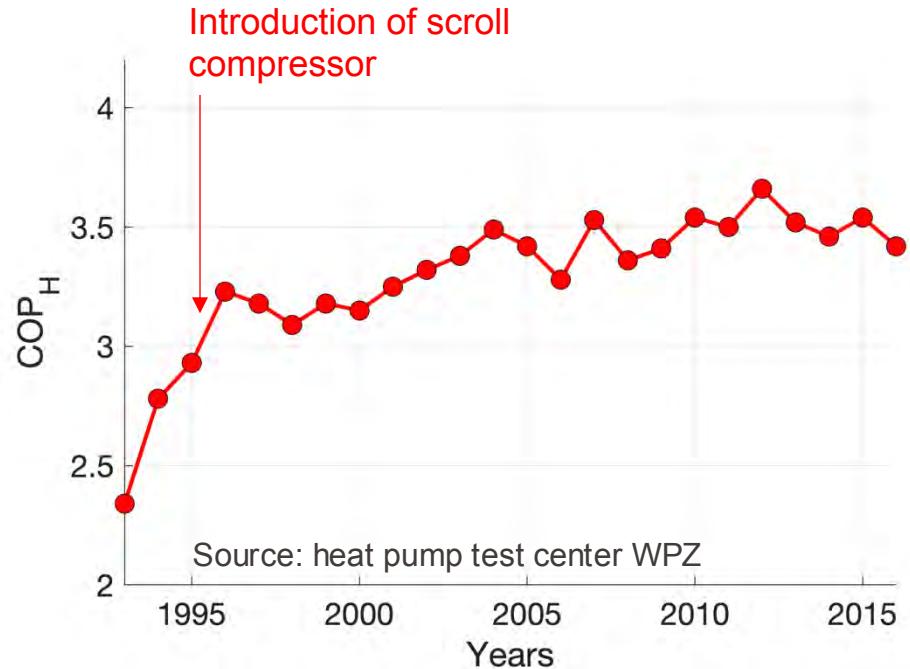
- 50% of required cuts achieved by increasing conversion efficiency
- 27% of primary energy goes into domestic heating and air-conditioning

Domestic Heating: Comparison of Technology Combinations



- Heat pumps play key role in reducing energy consumption and CO₂ emissions

Evolution of COP for Heat Pumps

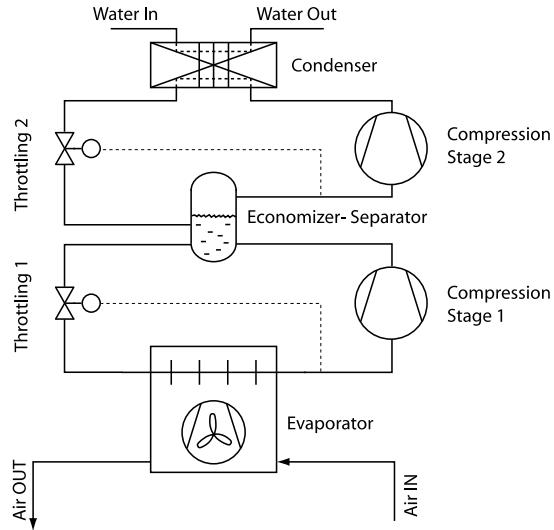


$$COP = \frac{E_{Heat}}{E_{Electricity}}$$

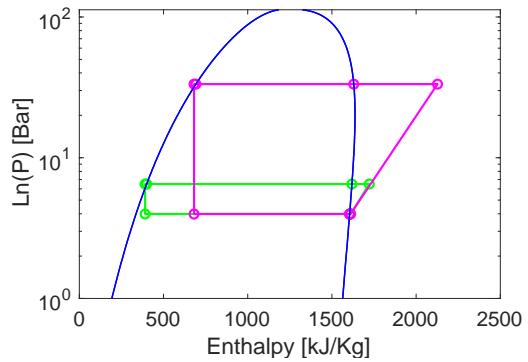
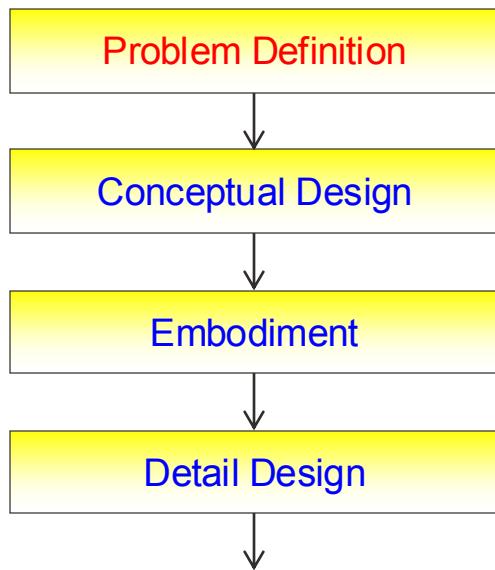
- Since introduction of scroll compressors, COP has been rising slowly
- Key question: can another step-change be achieved and if so, how?

Key Challenges

- Assessment of losses in heat pump cycles
 - Compression 50%
 - Expansion 30%
 - Heat transfer 20%
- Possible ways to reduce losses
 - Increase compressor efficiency
 - Use oil-free technology
 - Implement multistage cycles
- Potential enabler
 - Turbocompressors on oil-free bearings

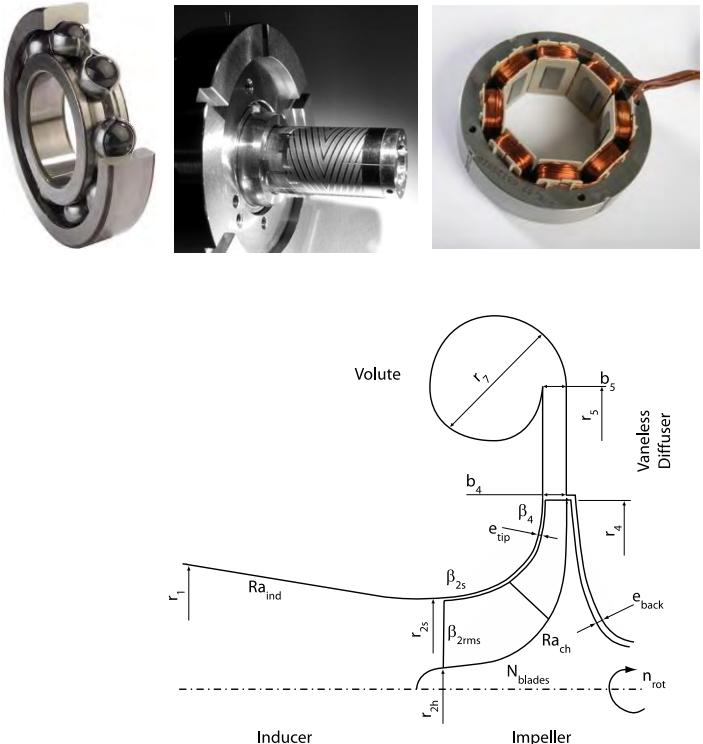
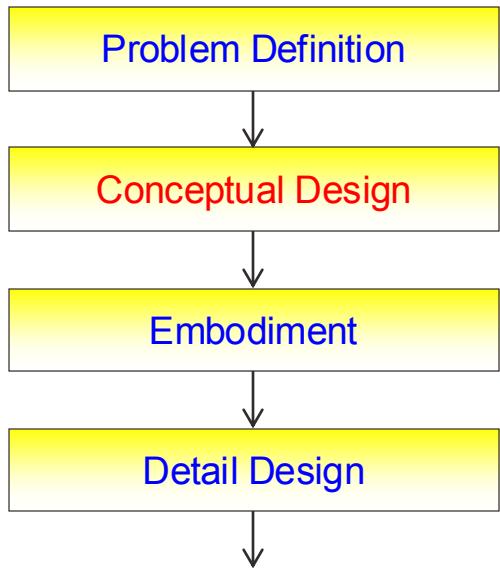


Designing an Oilfree Turbocompressor

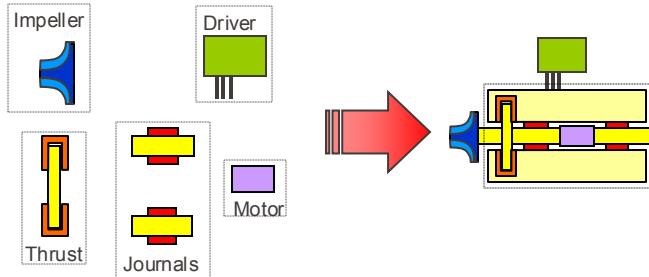
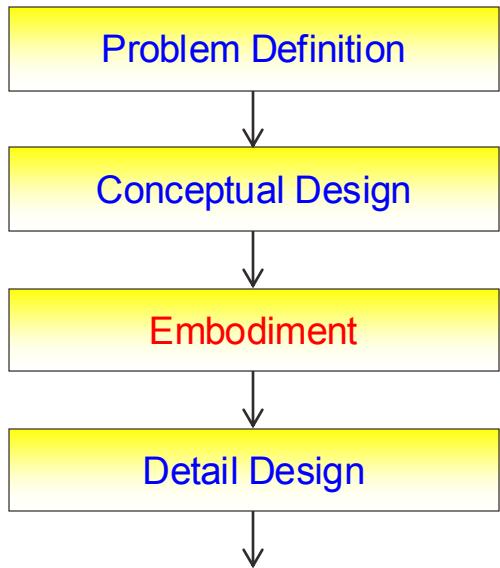


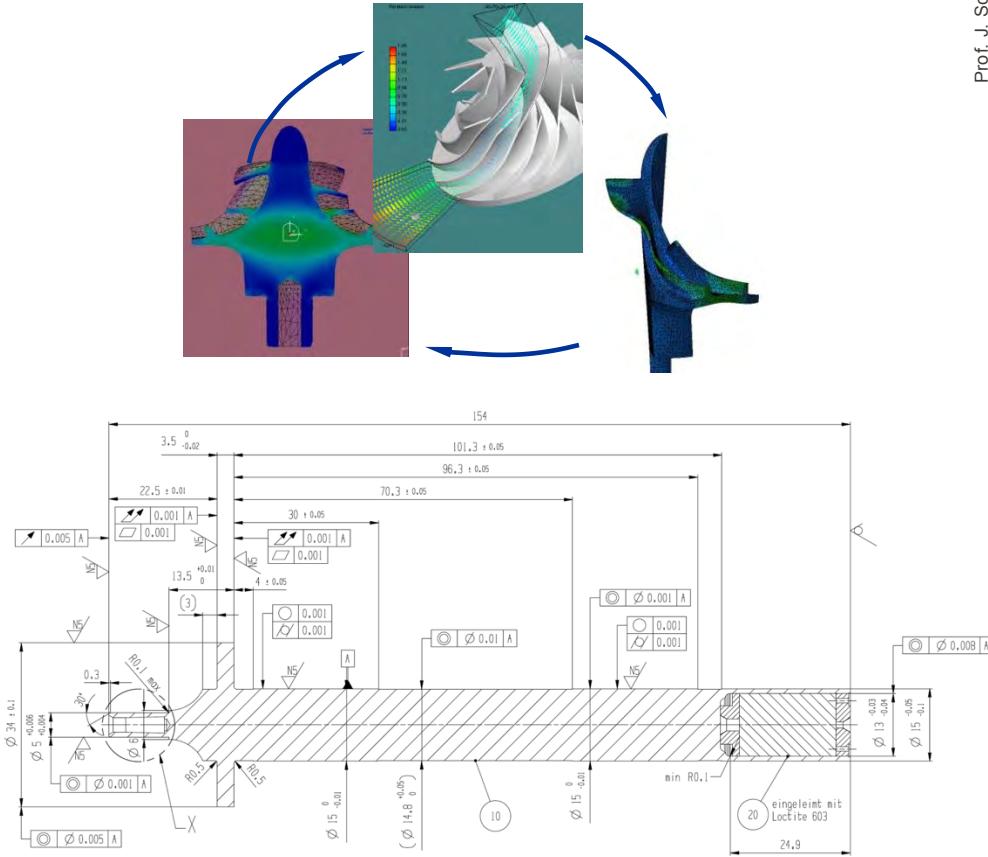
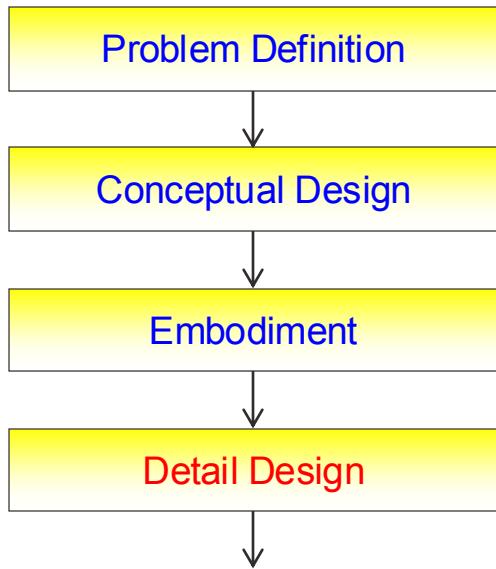
- Domestic heat pump
- Range of mass-flows, inlet T&P
- Working fluid
- Expected lifetime

Designing an Oilfree Turbocompressor



Designing an Oilfree Turbocompressor





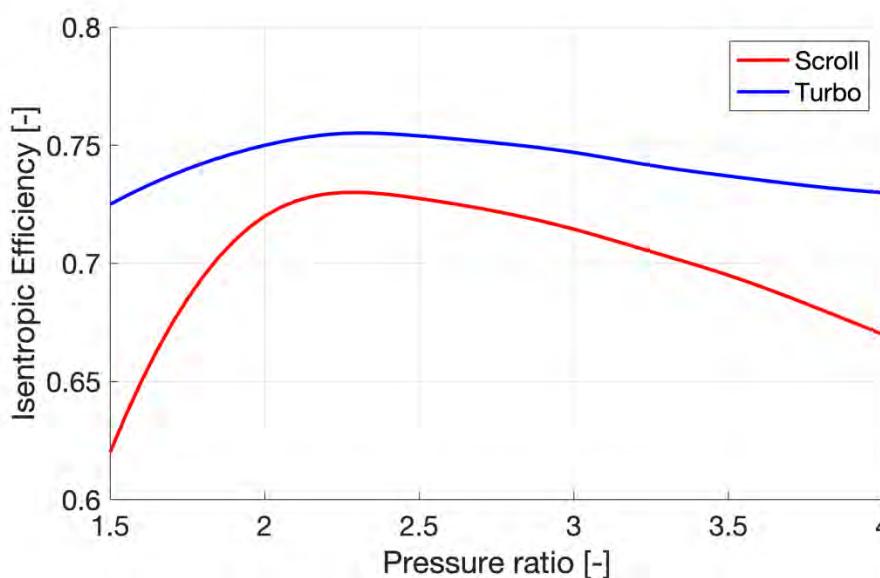
Proof of Concept Single Stage Compressor

- Experimental demonstration of small-scale turbocompressor
 - $\varnothing 20$ mm impeller, 210 krpm, 2 kW, Π 3.3
 - Oil-free, R134a-lubricated bearings
 - Increased specific power (x10)

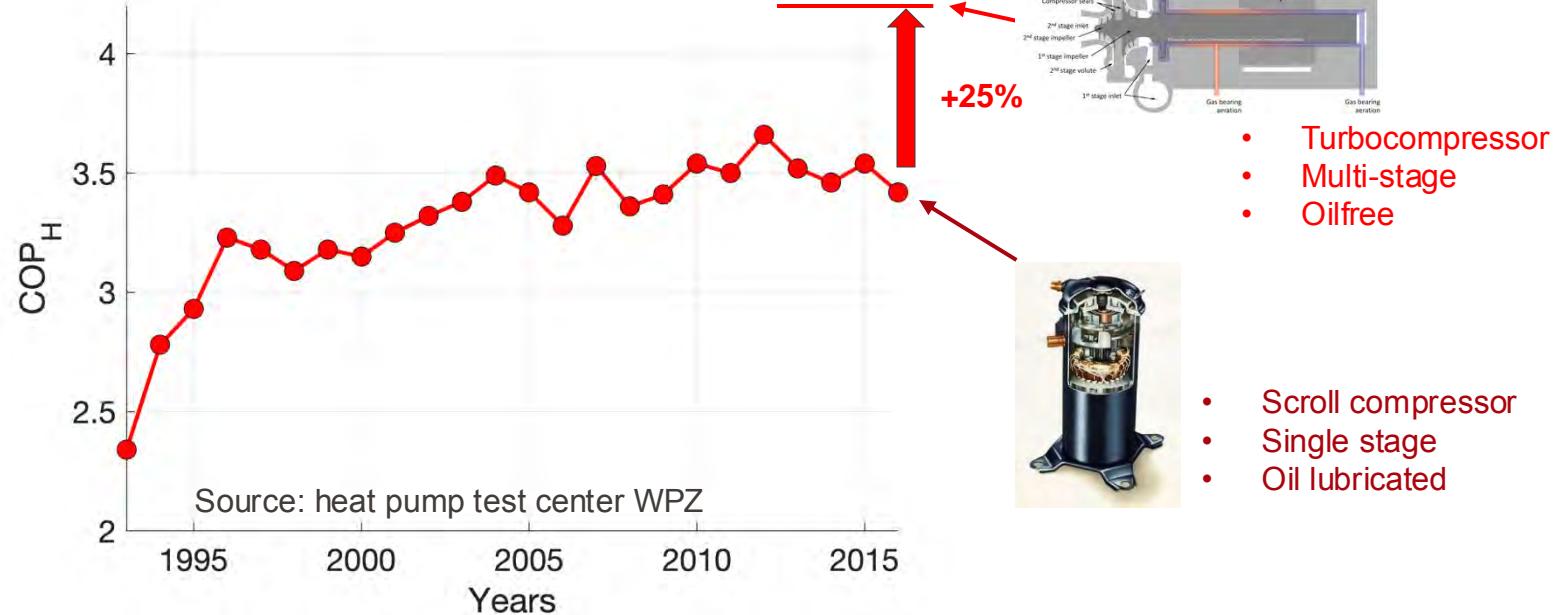


Proof of Concept Comparison

- Turbocompressor achieves higher peak efficiency than positive displacement compressor and improves off-design operation

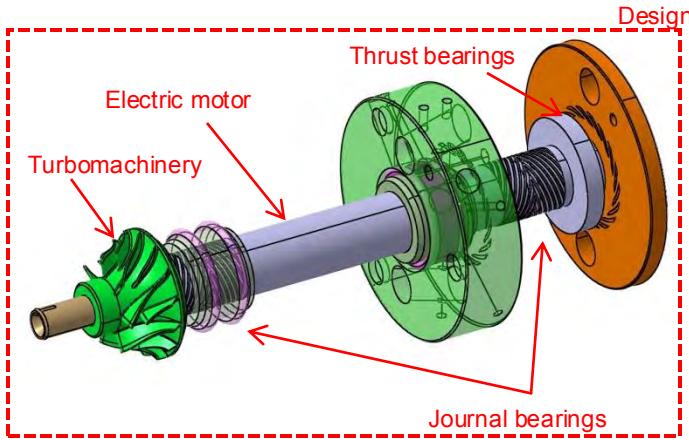


Potential of Oilfree Turbocompressors on COP



Design Challenges of High-Speed Turbomachinery

- Pronounced multidisciplinarity
- Strong interactions between components
- Competing design objectives
- High level of design constraints
- Traditional design approach based on fragmented component view
 - Tends to neglect component interactions
 - Hindrance to novel solutions
 - Yields suboptimal solutions
 - Manual iterations leave no peace of mind

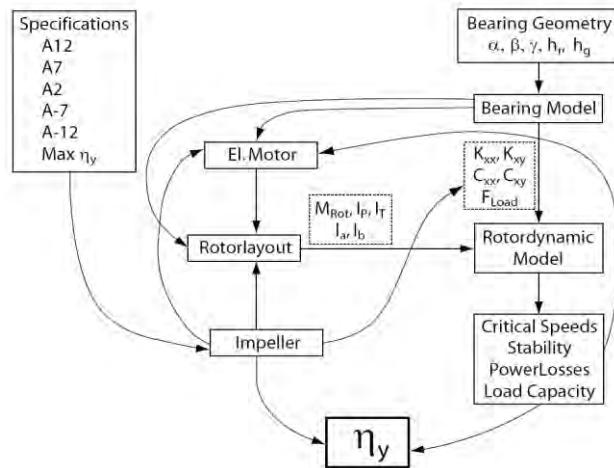


- Design problem formulated as multi-objective optimization

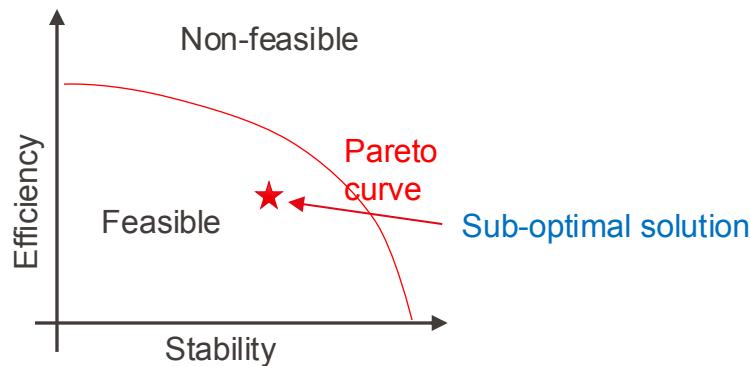
$$\min f(G) = -[\eta_{is}(G), \Gamma(G)]^T$$

- Subject to constraints

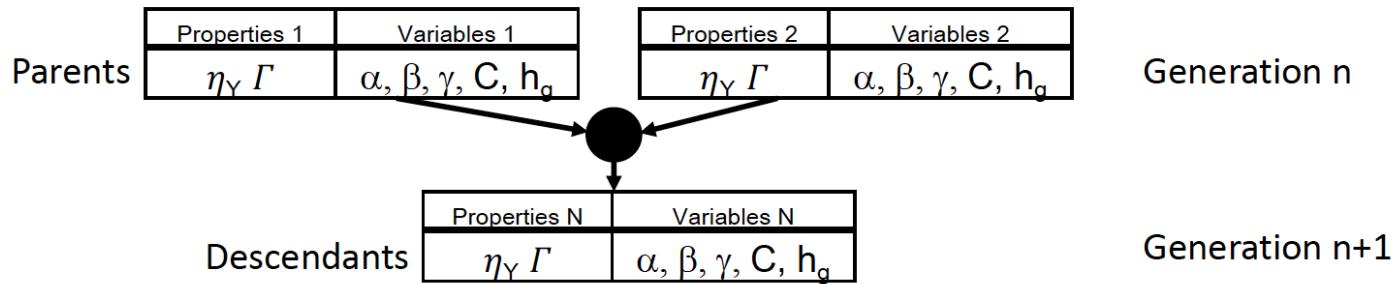
- Bearing clearance
- Bearing load capacity at low speed
- Thrust bearing load capacity
- Margin to 1st lateral bending eigenmode
- Avoidance of compressor surge
- Mechanical stress below threshold

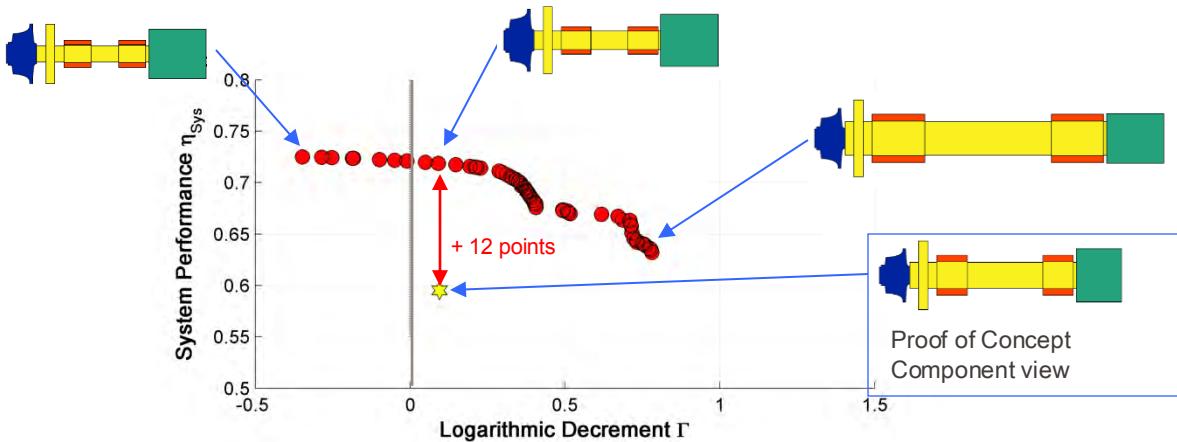


- Competing objectives result in Pareto curve
 - Pareto curve splits objective domain into feasible and non-feasible regions
 - Pareto curve is a family of non-dominated solutions
 - Selecting “right” objectives can be challenging



- Evolutionary algorithms
 - Procedure is started with initial population of solutions
 - Each individual is characterized by genes (variables) and by resulting performance (indicators)
 - Individuals are crossed and descendants generated
 - Fit solutions are kept, unfit ones discarded

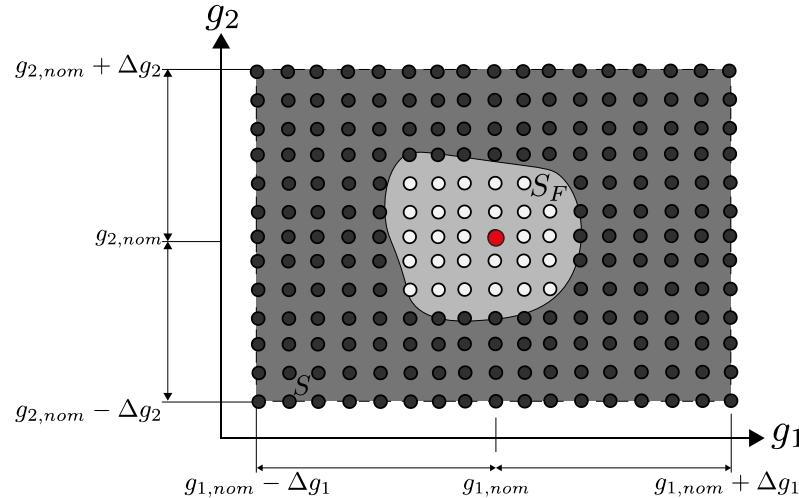




- System approach captures complex interactions and improves compressor performance by 12 points compared to component view
- Are these "right" objectives? What about robustness?

How to Define Design Robustness?

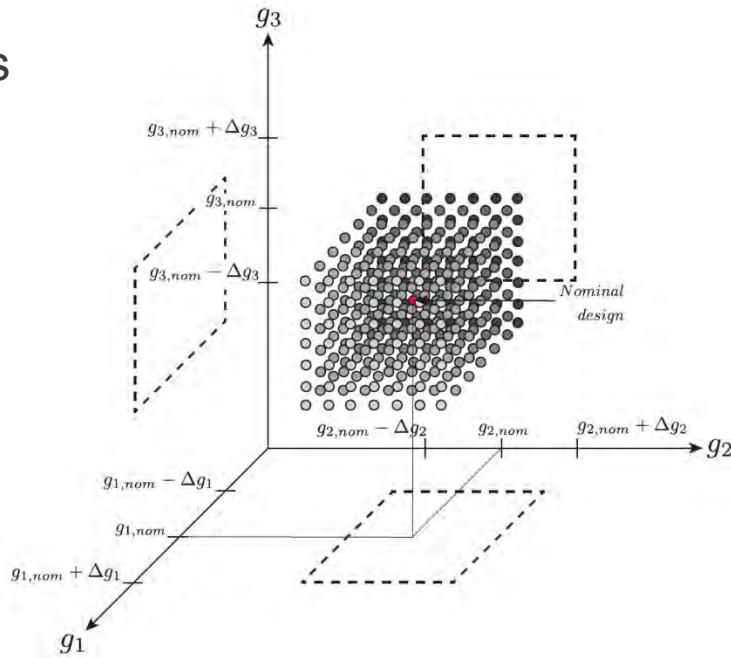
- Maximize hypervolume defined by feasible design within assessed deviation volume around nominal design
- Maximize signal-to-noise ratio to maximize performance and reduce variance across hypervolume



$$S/N_f = 10 \cdot \log_{10} \left(\frac{\mu_f^2}{\sigma_f^2} \right)$$

Challenges to Identify Robust Designs

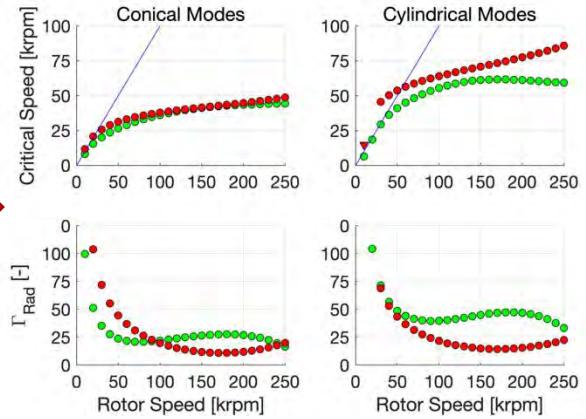
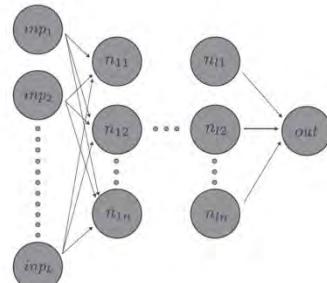
- Increased number of objectives
- Increased number of evaluations to assess effect of deviation requires fast models
- Need fast and accurate surrogate models



Neural Network Based Surrogate Model

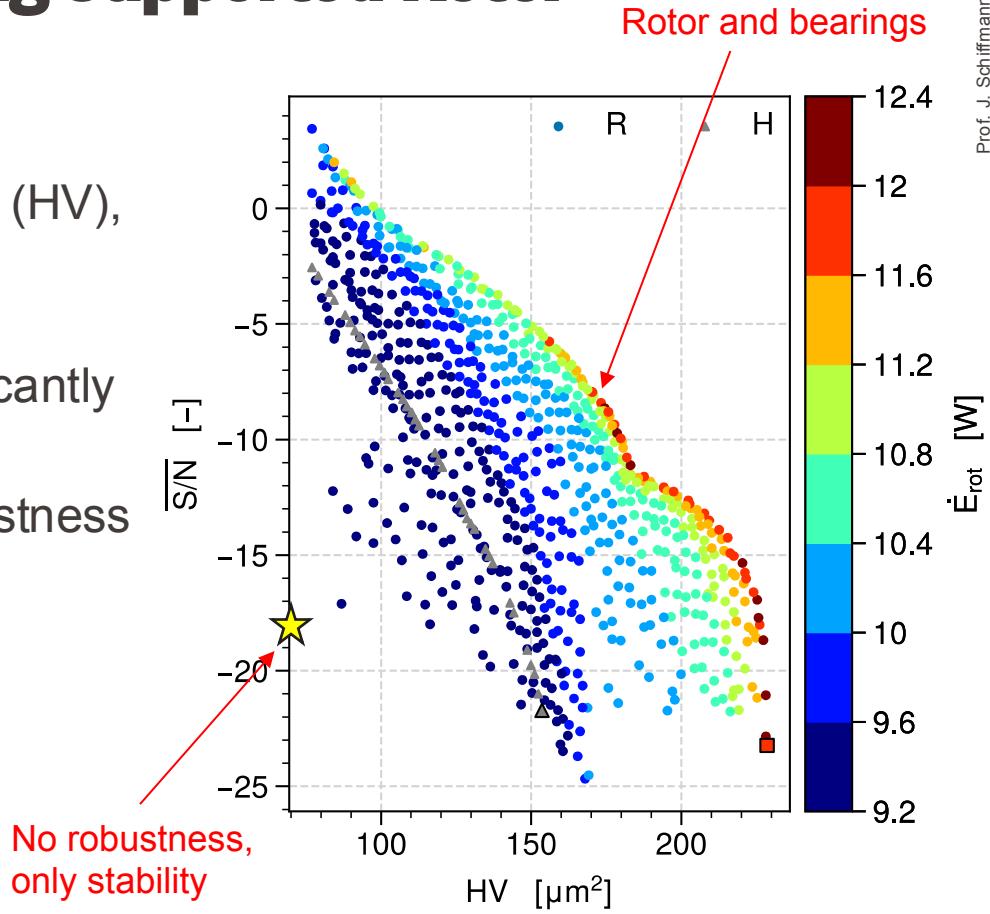
- Input is bearing and rotor geometry, output is whirl speed map
- Relative error <3% between high-fidelity and surrogate model
- Surrogate model based on ensemble of networks

Variable	Symbol
Groove-ridge width ratio	$\alpha = \frac{a}{a+b}$
Groove angle	β
Grooved region ratio	$\gamma = \frac{L_o - L_{land}}{L_o}$
Bearing length to diameter ratio	$LoD = \frac{L_o}{2R}$
Ratio of bearings distances to CG	$H_{gr} = 1 + \frac{h_a}{h_r}$
Mass number	$\bar{M} = \frac{Mh_r\Omega^2}{P_aR^2}$
Transverse moment of inertia number	$\bar{I}_t = \frac{I_t h_r \Omega^2}{P_a R^2 L_o^2}$
Polar moment of inertia number	$\bar{I}_p = \frac{I_p h_r \Omega^2}{P_a R^2 L_o^2}$
Bearing A distance to CG	$\bar{L}_A = \frac{L_A}{L_o}$
Bearing B distance to CG	$\bar{L}_B = \frac{L_B}{L_o}$
Compressibility number	$\Lambda = \frac{6\mu\Omega \cdot (D/2)^2}{P_a h_r^2}$

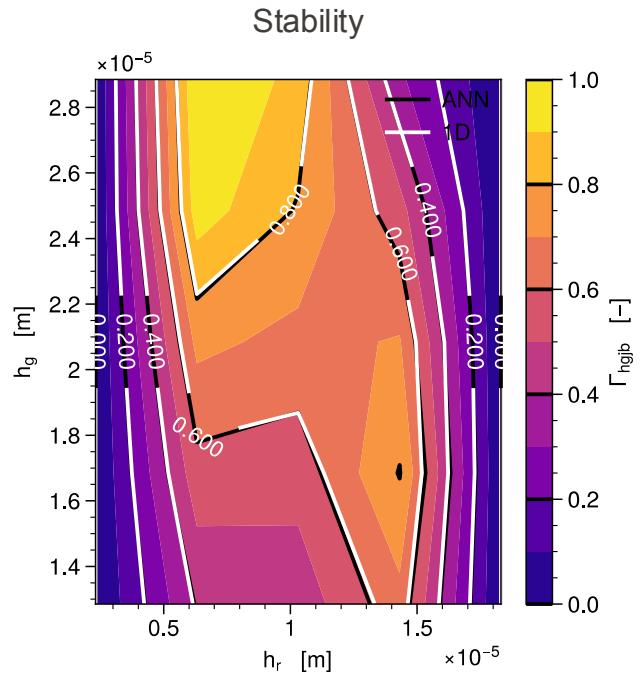


Test Case: Gas Bearing Supported Rotor

- Tradeoff between robustness (HV), losses, and variance (S/N)
- Robustness increased significantly
- Supports idea of design robustness
- Increased robustness offers cost savings

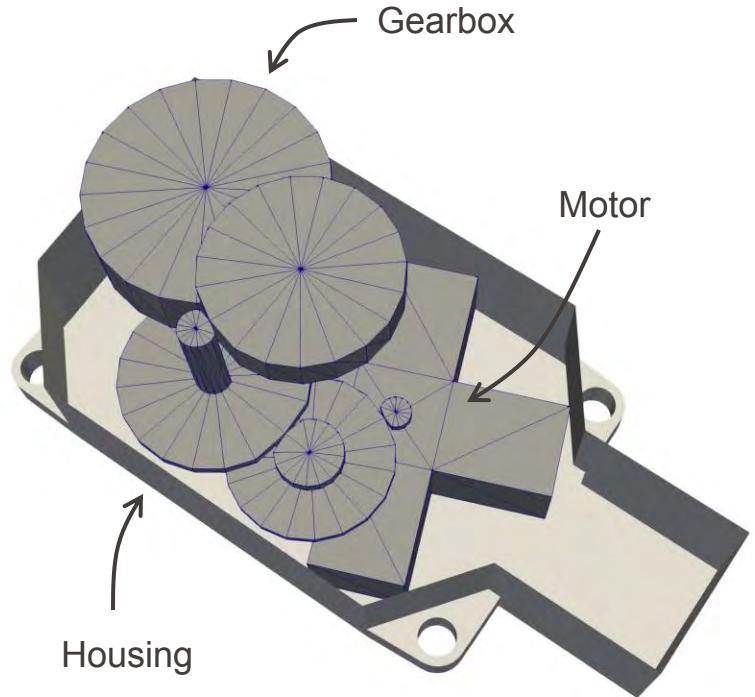
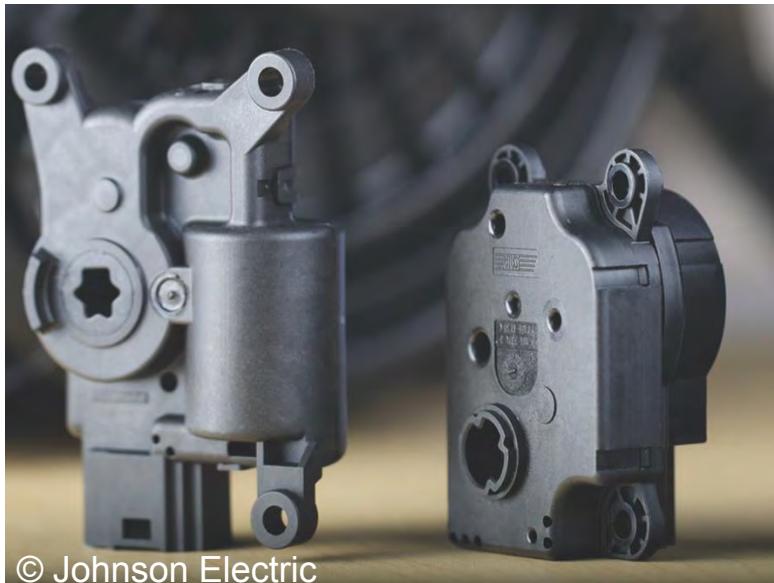


Test Case: Deviation Effects on Objectives



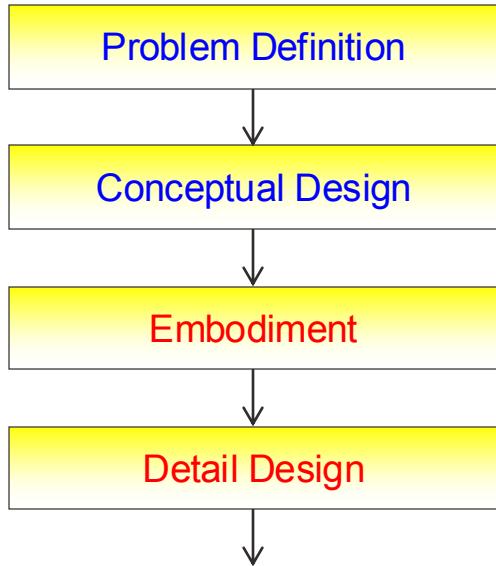
- Stable operation with radial clearance $10\pm8\text{ }\mu\text{m}$ instead of $7\pm1.5\text{ }\mu\text{m}$
- Excellent agreement between high-fidelity and surrogate model
- Baseline model optimization performed on HPC of EPFL
 - 96'000 CPU-core-h, € 1'300, 750kWh
- Surrogate model based optimization performed on GPU
 - 1.5 GPU-h, € 1, 0.5kWh

Design of Electromechanical Actuators

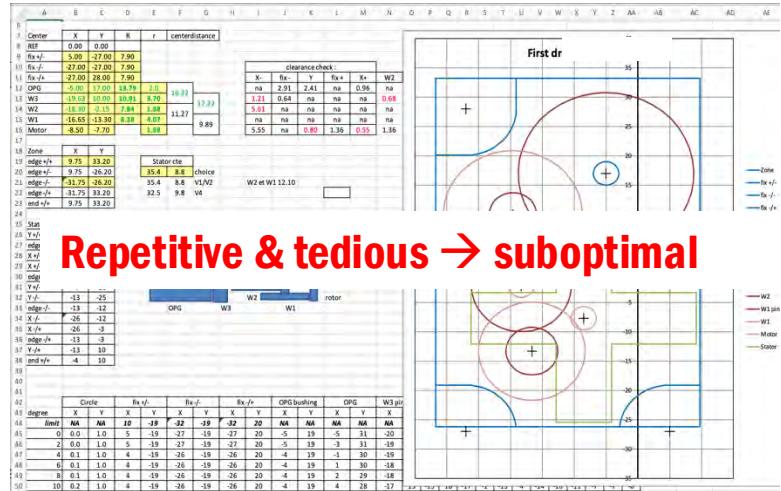


- Actuators for car industry (ventilation, light, grill, parking brake)

- Embodiment & Detailed design

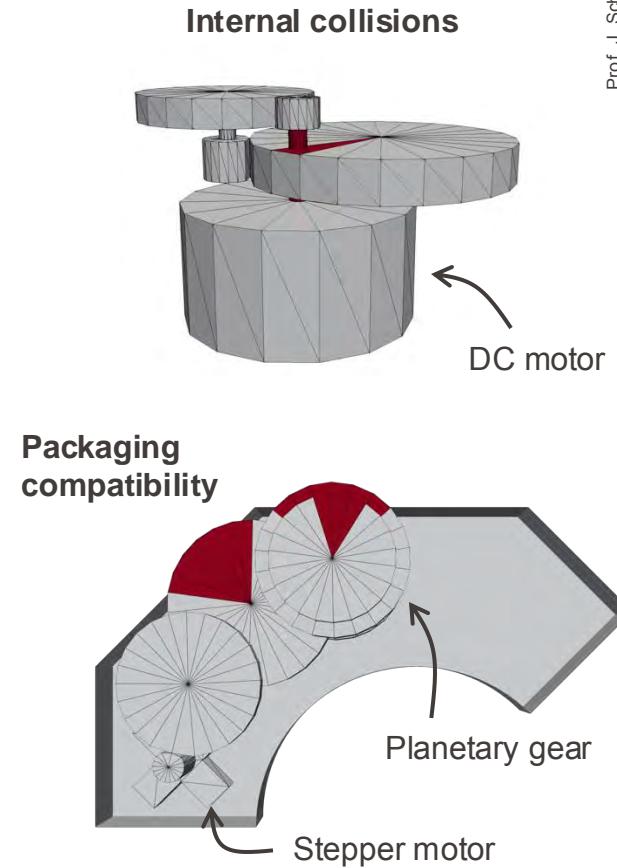
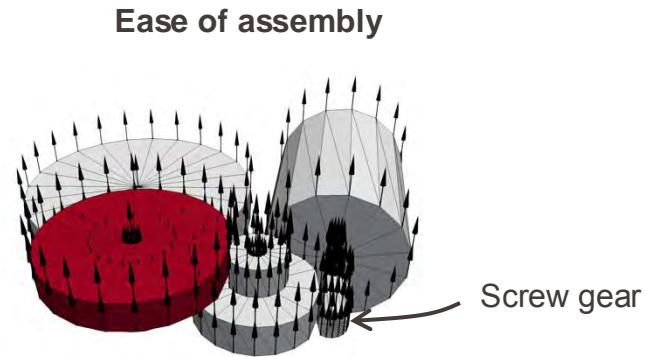


- Routine task with manual optimization
- 1 project manager and 2 specialists working ~ 1 month

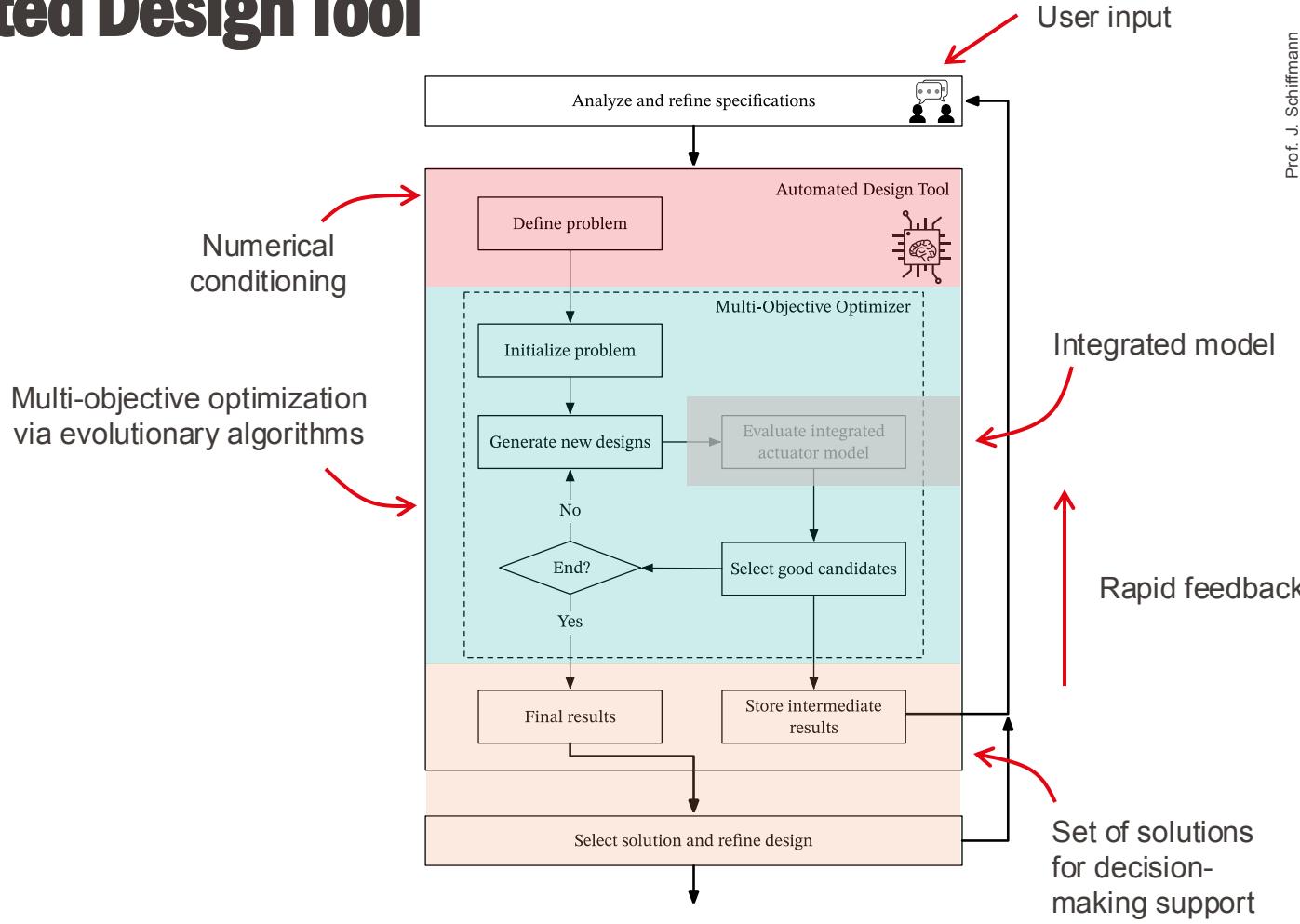


Integrated Actuator Model

- Modular components
 - Physics (motor, gearbox)
 - Cost
- 3-D mesh enables identification of constraints
 - Triangle-triangle collision detection
 - Ray-tracing algorithms

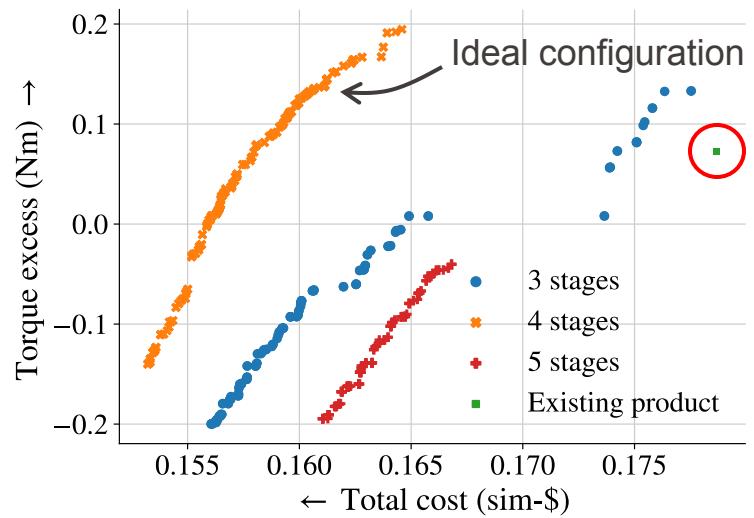


Automated Design Tool



Example: Actuator for Ventilation Flaps

- Optimization for max torque and min cost
- Obtained solutions outperform solution by experienced engineers
- Pareto optimal solution obtained within < 2h on classical laptop



- Push design automation towards earlier phases of process
- Refine surrogate modeling techniques
- Alternative optimization methodologies
- Computationally cheap digital-twins for real-time monitoring



Thank you for your attention

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