Optimization and Robustness Evaluation of an Axial Turbine using Fluid-Structure Interaction

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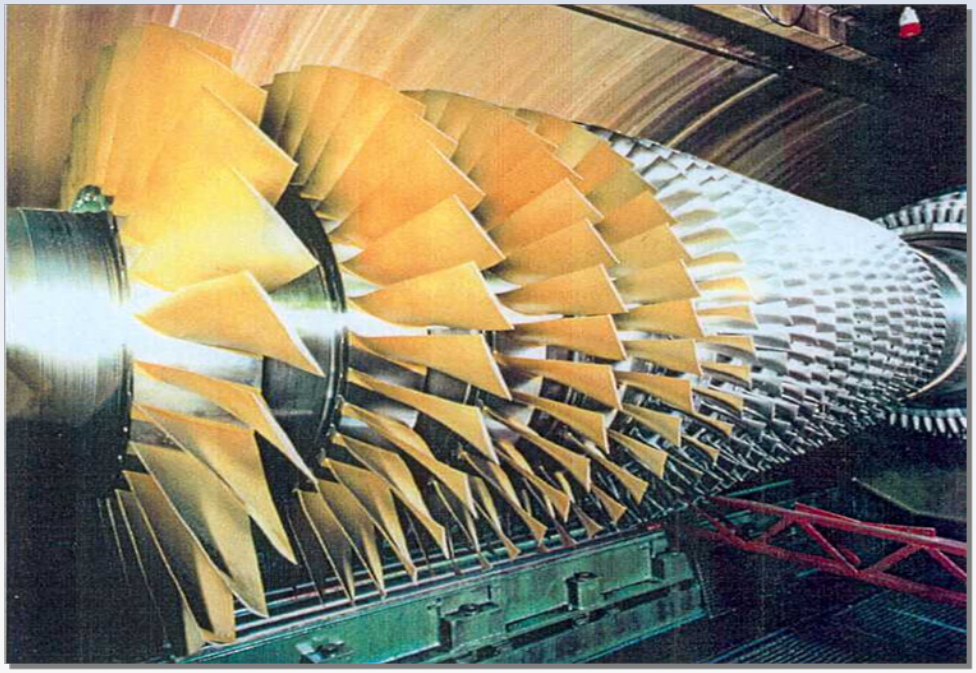
Outline

- **Parametric Process Integration**
- **Sensitivity Analysis**
- **Design Optimization**
  - Evolutionary Algorithm
  - Adaptive Response Surface Method
- **Robustness Evaluation**
- **Outlook**
  - Random Fields
  - Design for Six Sigma (Reliability Analysis)

Motivation



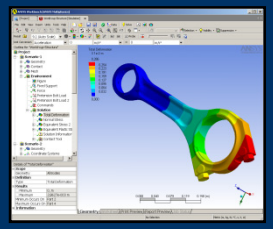
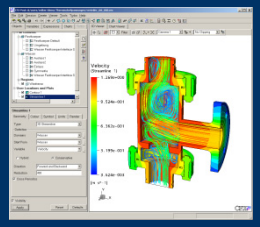
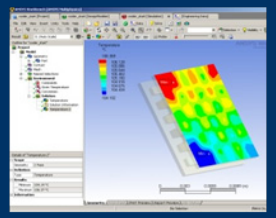
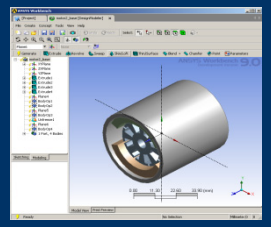
Power Plant	1000 MW
Efficiency	50 %
Increase of 1%	+20 MW
=Electricity for	120 000 Inhabitants

Workbench Platform & optiSLang

CAD / PDM

ANSYS Workbench

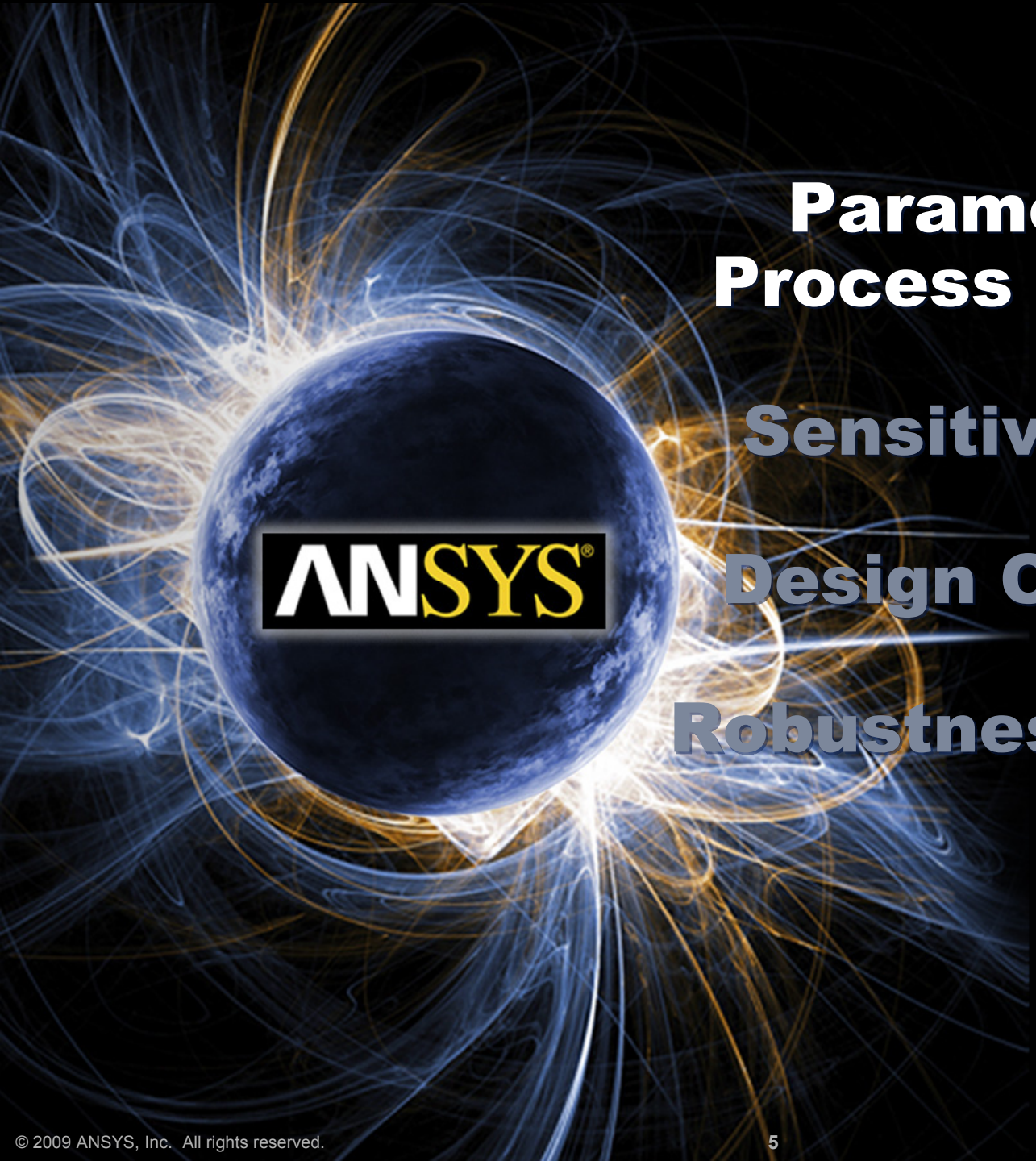
Structural Mechanics - Fluid Dynamics - Heat Transfer - Electromagnetic

A Multi-Physics Design and Analysis System

Sensitivity | Optimization | Robustness | Reliability | Robust Design

optiSLang



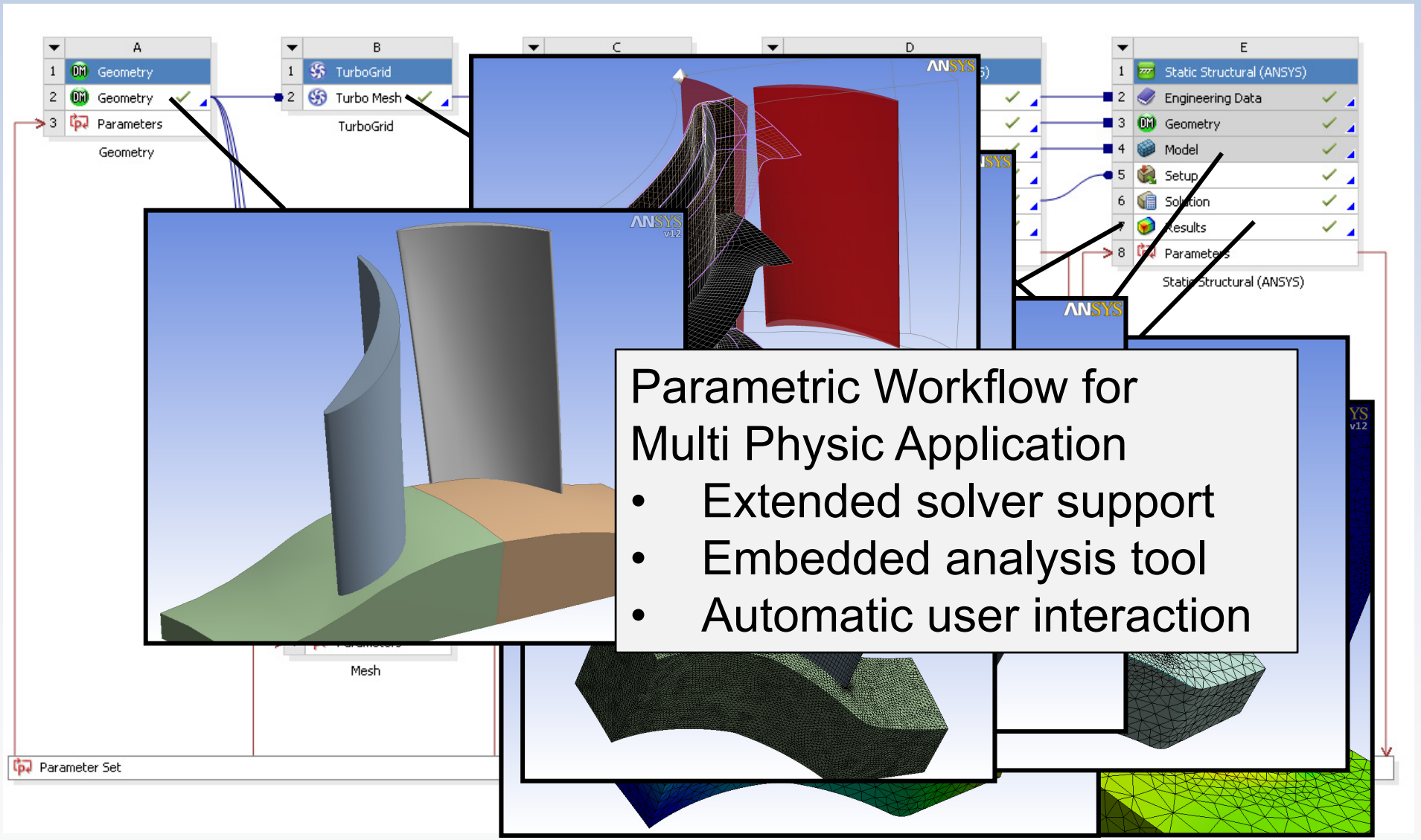
Parameterization
Process & Geometry

Sensitivity Analysis

Design Optimization

Robustness Evaluation

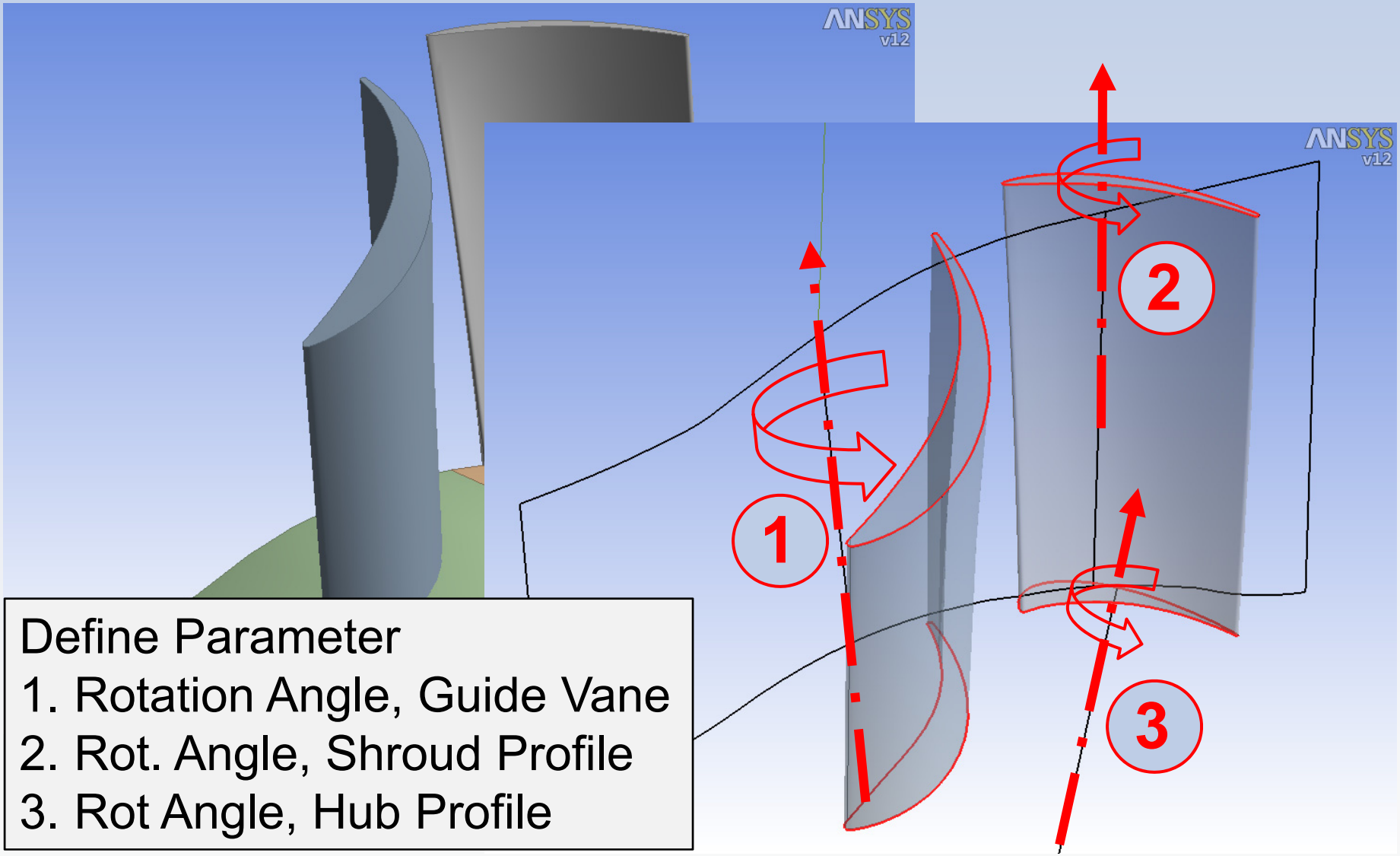
Parameterization of the Workflow



Parametric Workflow for Multi Physic Application

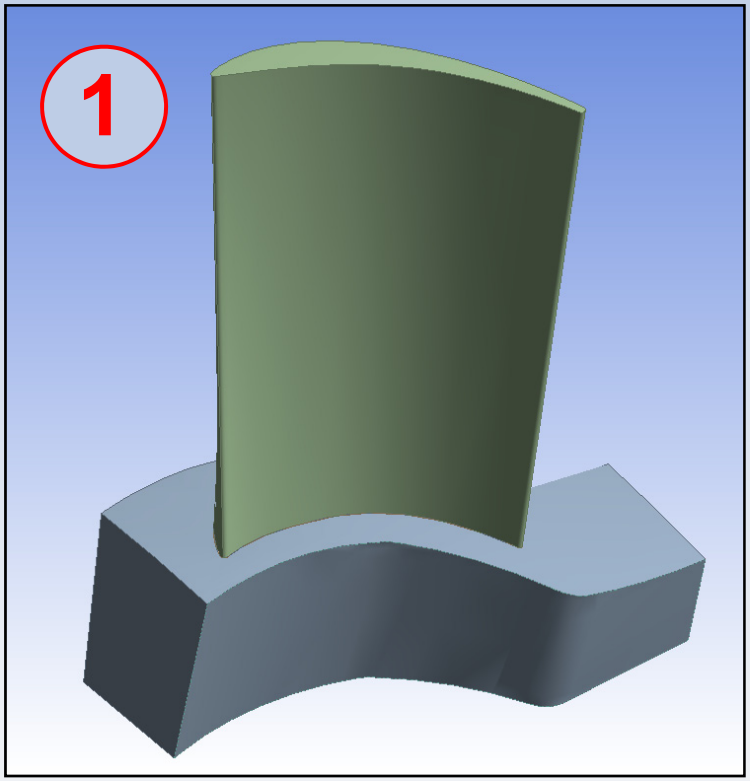
- Extended solver support
- Embedded analysis tool
- Automatic user interaction

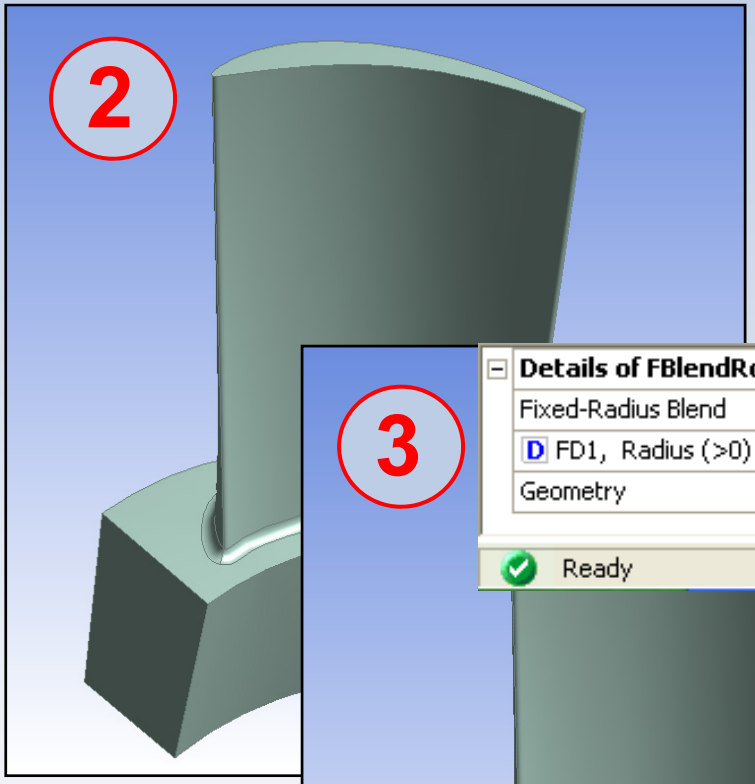
Parameterization of the Geometry

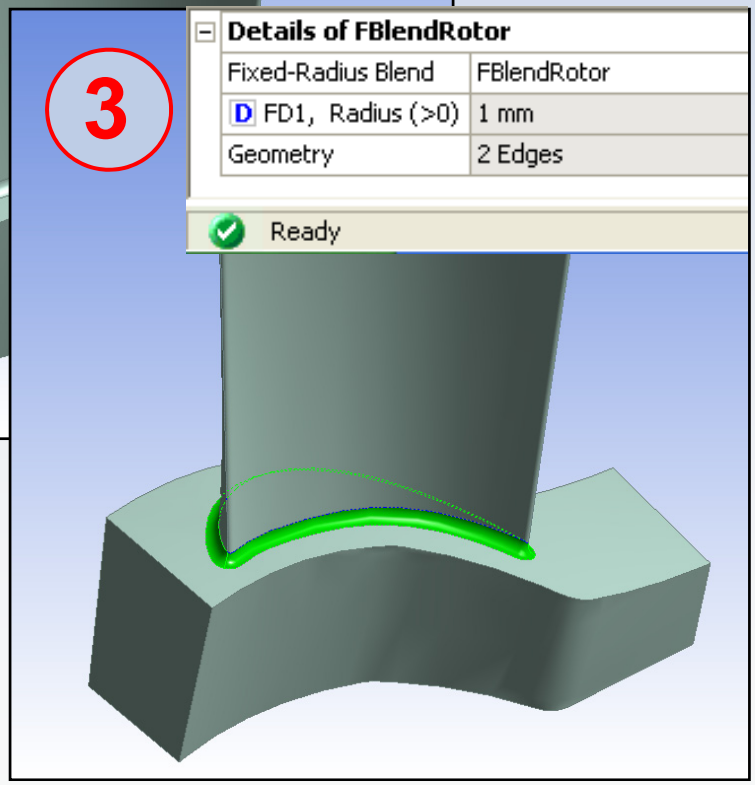


Define Parameter
1. Rotation Angle, Guide Vane
2. Rot. Angle, Shroud Profile
3. Rot Angle, Hub Profile

Parameterization of the Geometry

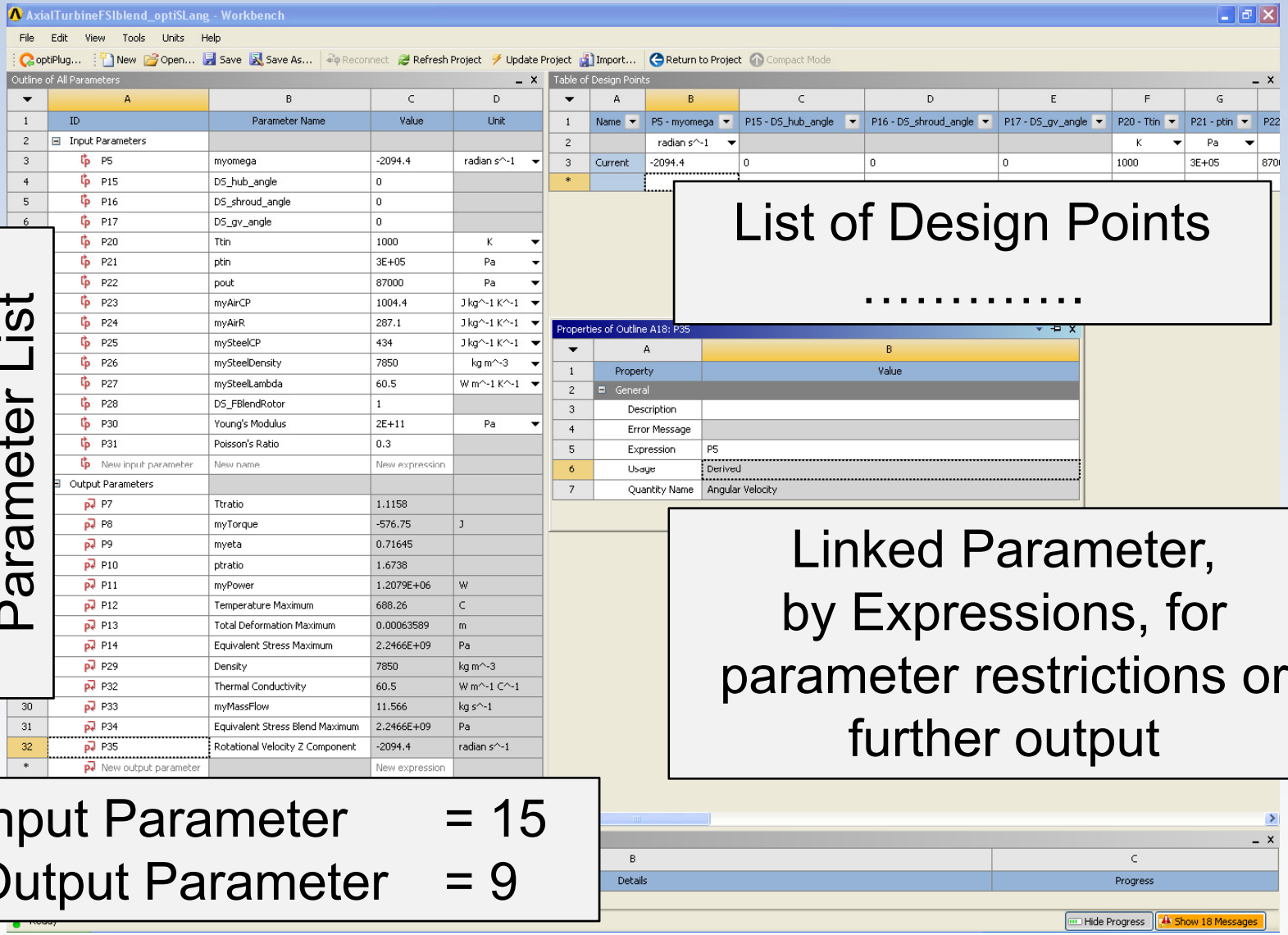




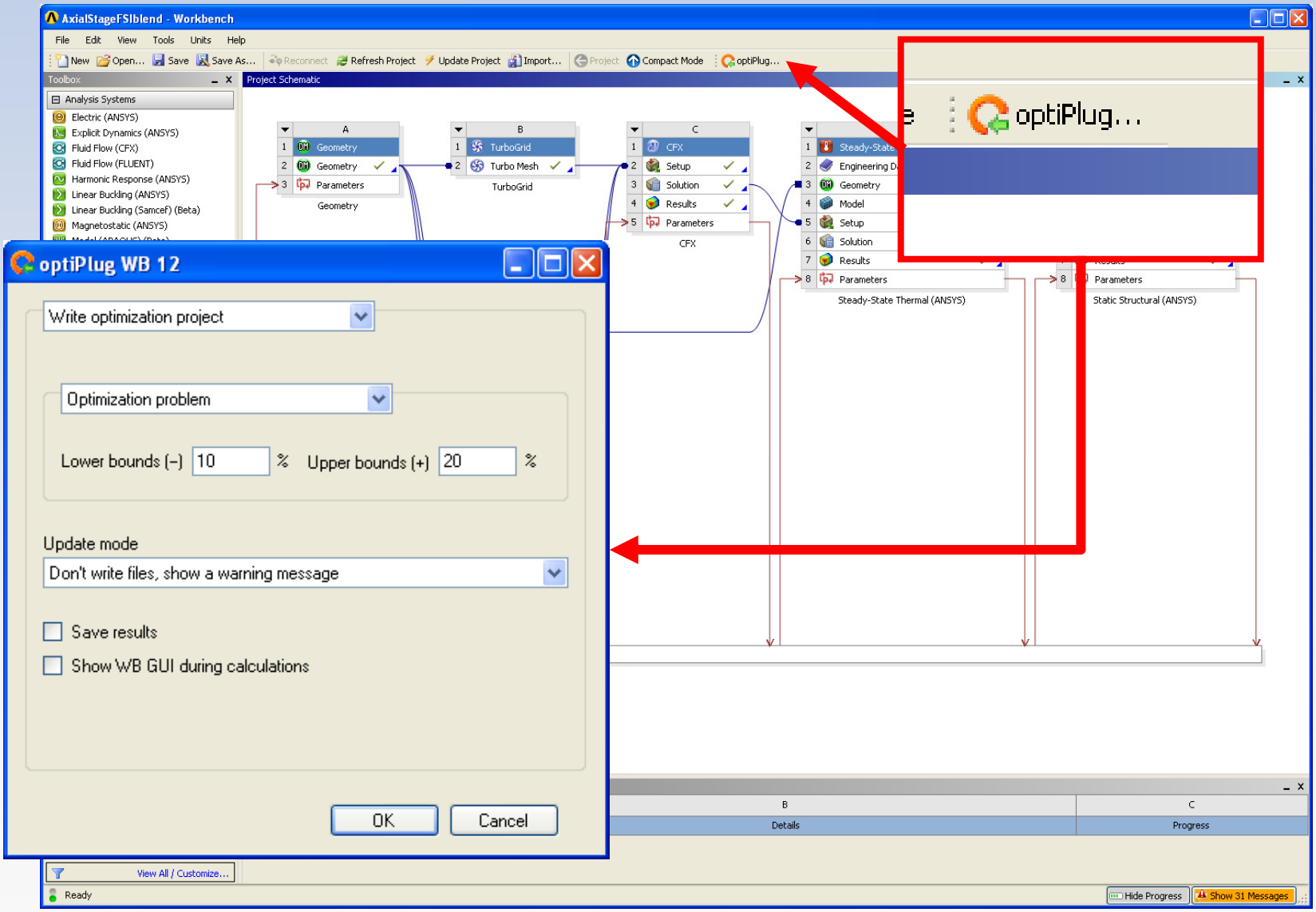


1. CHT Model: Multi Body Part
2. CSD Model: Single Part
3. Add Blend for CSD, as Parameter

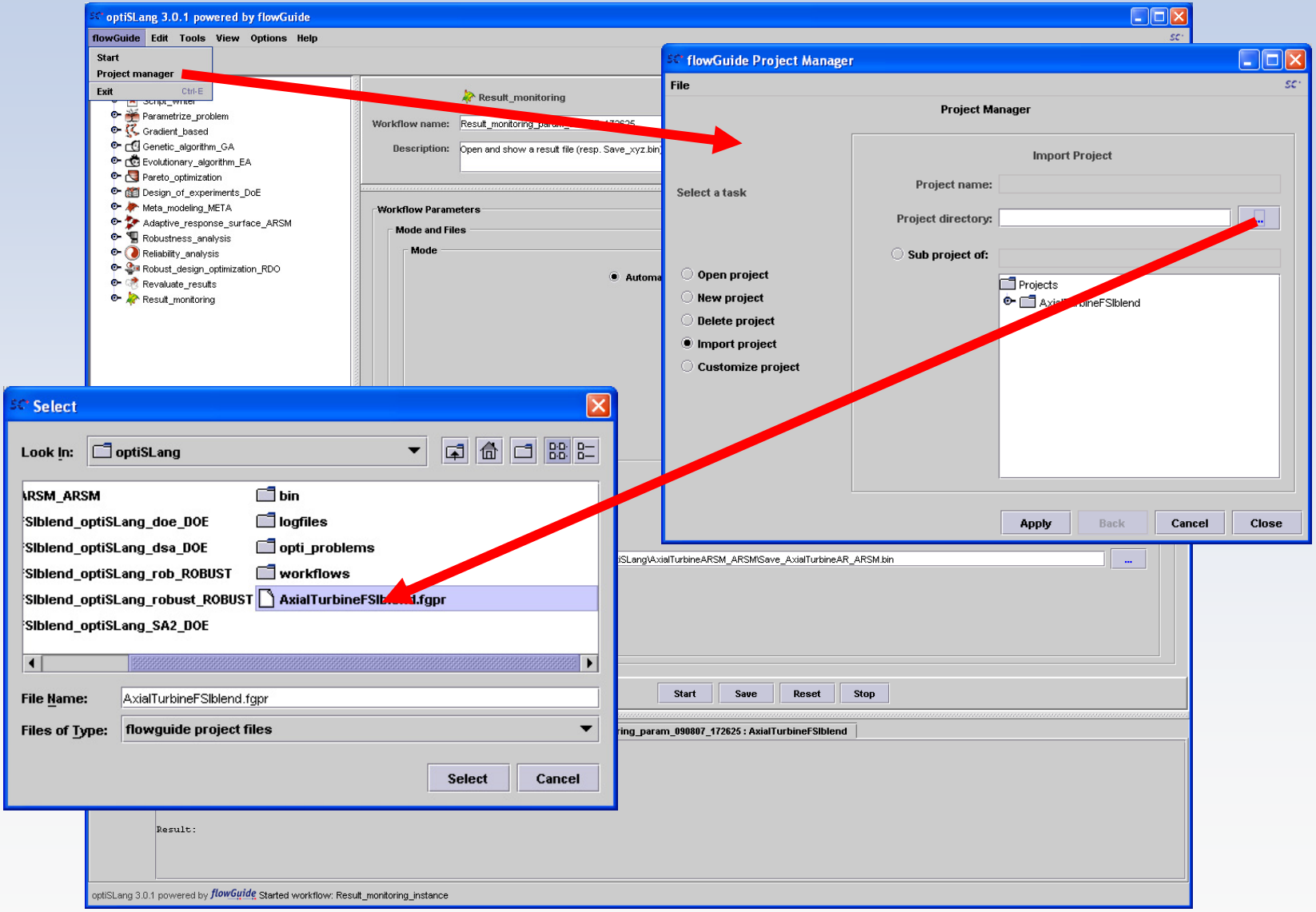
Parameter Manager



Workbench Interface to optiSLang



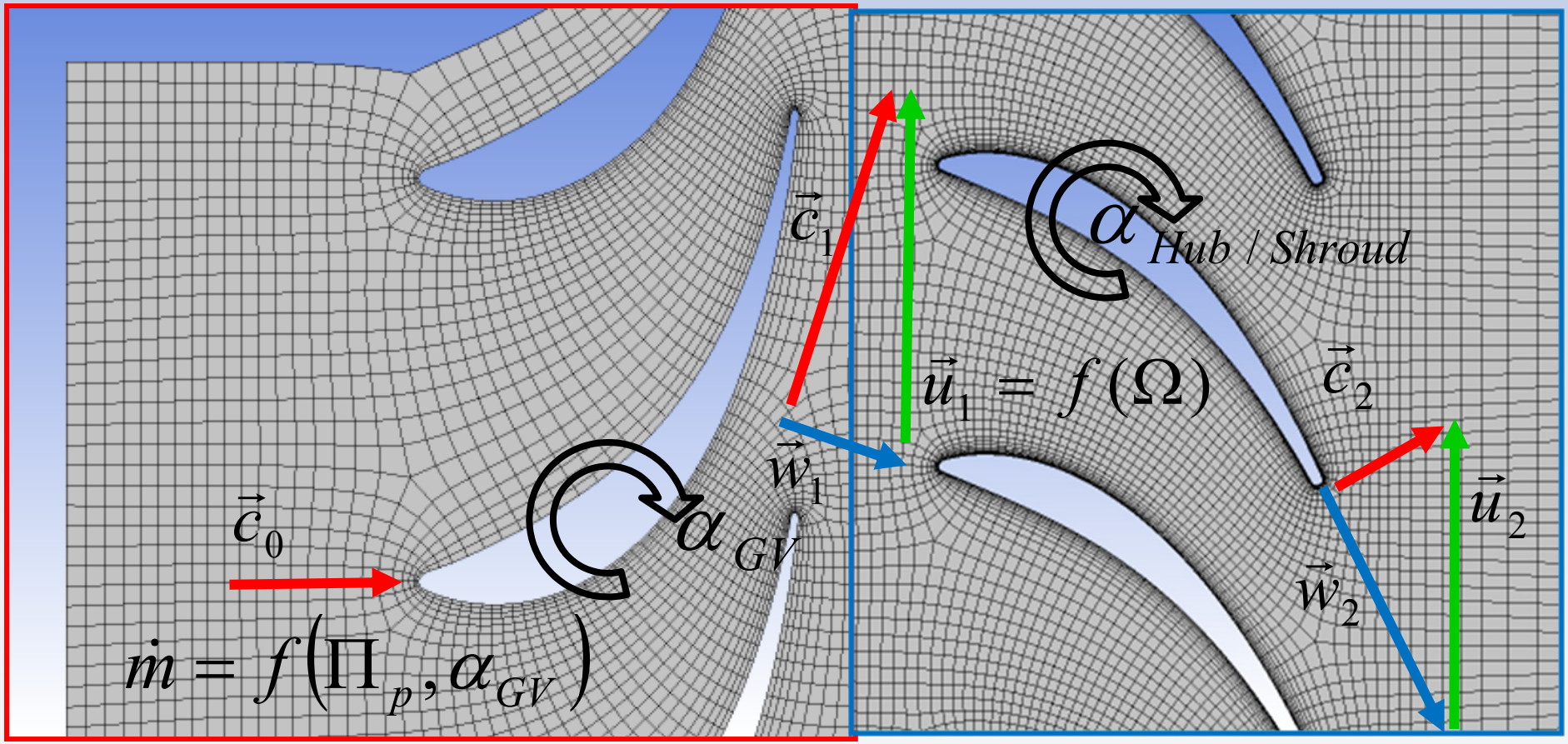
Workbench Interface to optiSLang



Parameter Attributes

Input Parameter	Parameter Name	Initial Value	Type
Blade Angels	$\alpha_{GV}, \alpha_{Hub}, \alpha_{Shroud}$	0°, 0°, 0°	deterministic
Rotational Velocity of Rotor	Ω	-2094 [rad/s]	deterministic
Rotor Blend Radius	r_{Blend}	1 [mm]	deterministic
Total Temperature Inlet	$T_{t,Inlet}$	1000 [K]	deterministic
Total Pressure Inlet	$p_{t,Inlet}$	400 [kPa]	deterministic
Pressure Outlet	p_{out}	187 [kPa]	stochastic
All Material Properties	-	-	stochastic
Output Parameter	**Parameter Name**	**Initial Value**	**Target**
Total Temperature Ratio	$\Theta_T = T_{t,Inlet}/T_{t,Outlet}$	1.115	-
Total Pressure Ratio	$\Pi_p = p_{t,Inlet}/p_{t,Outlet}$	1.673	-
Torque/Power at Rotor	M_P, P	-577 [Nm], 1.21 [MW]	maximize
Mass Flow Rate	m	11.56 [kg/s]	-
Isentropic Efficiency	η	71.64 [%]	maximize
Maximal v. Mises Stress	σ_{max}	218.6 [MPa]	below limit

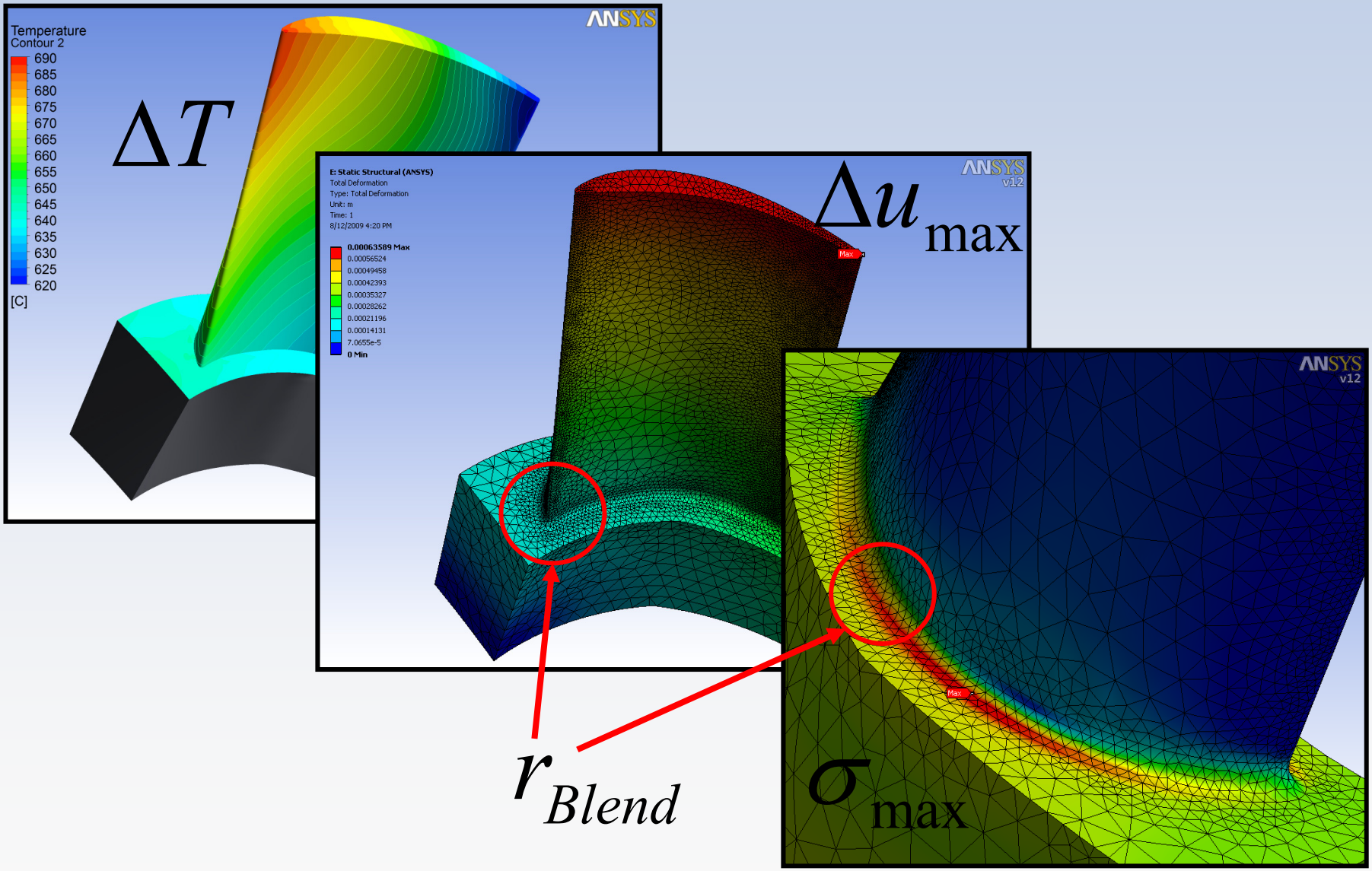
Parameter Physics, Fluid Flow

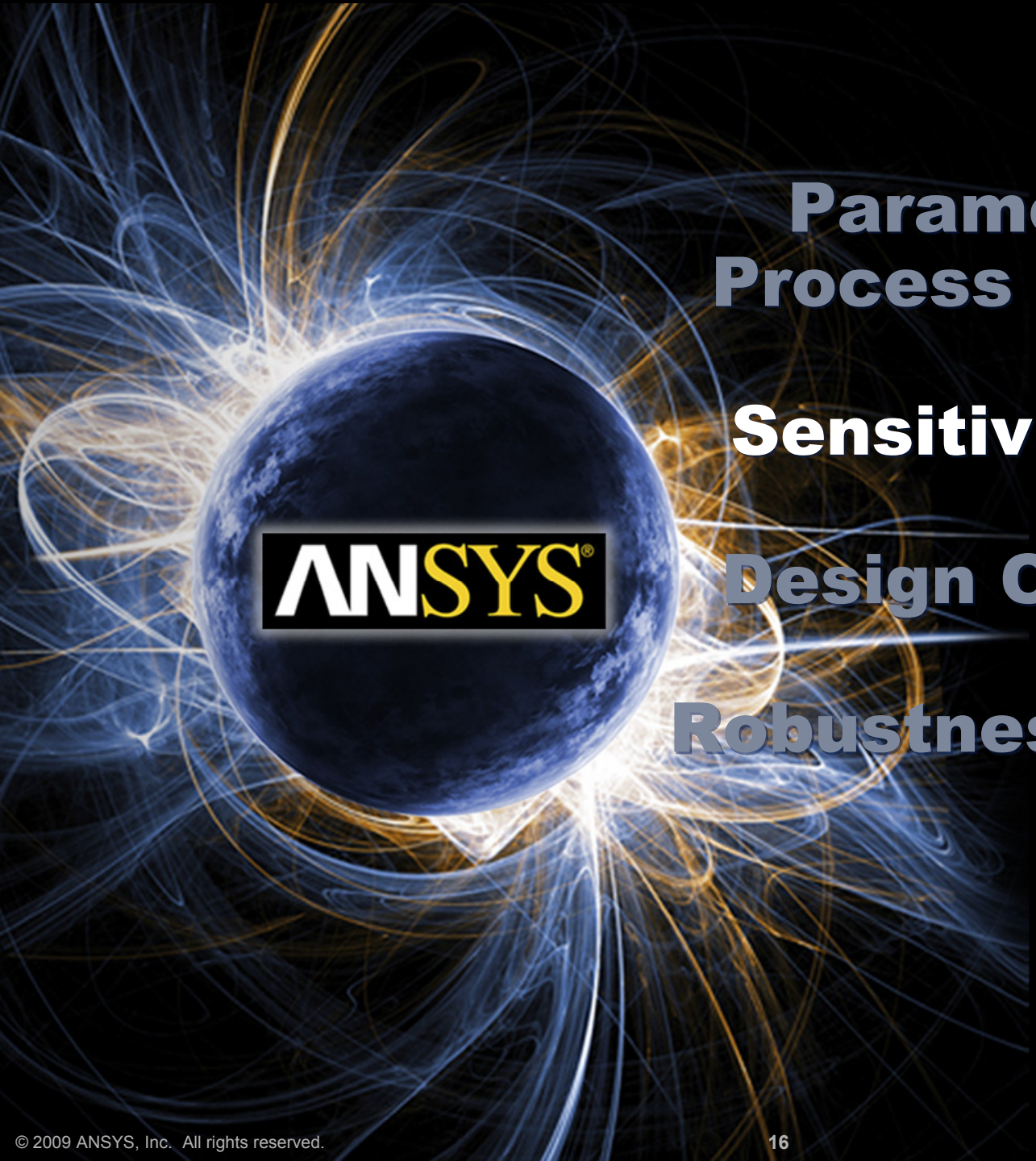


$$\eta = f\left(\Theta_T, \Pi_p\right)$$

$$P, M_P \sim \dot{m} \cdot \underset{1-2}{\Delta}\left(u \cdot c_u\right)$$

Parameter Physics, Mechanic





Parameterization
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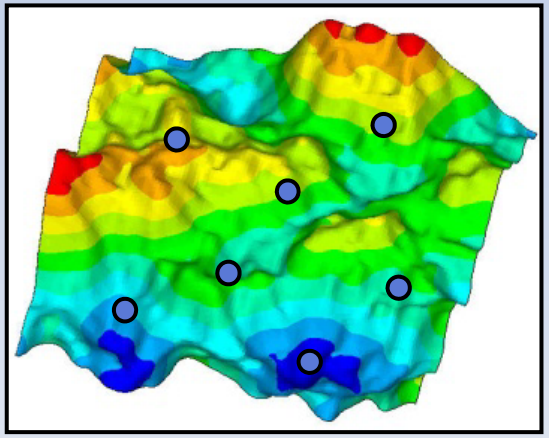
Sensitivity Analysis

Design Optimization

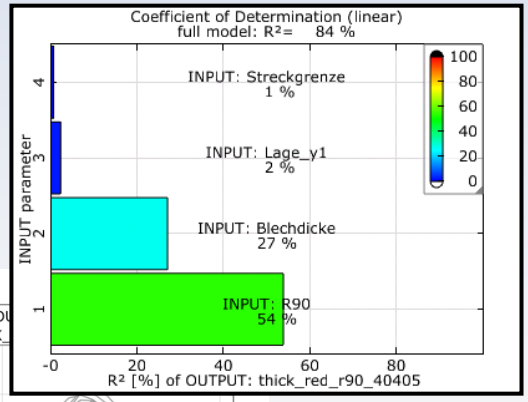
Robustness Evaluation

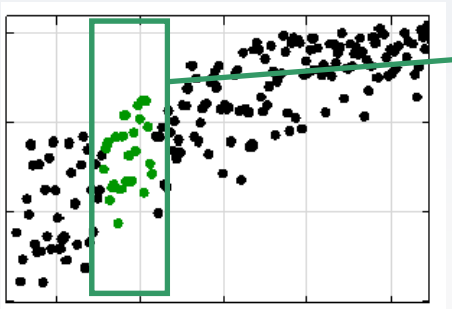
Sensitivity Analysis

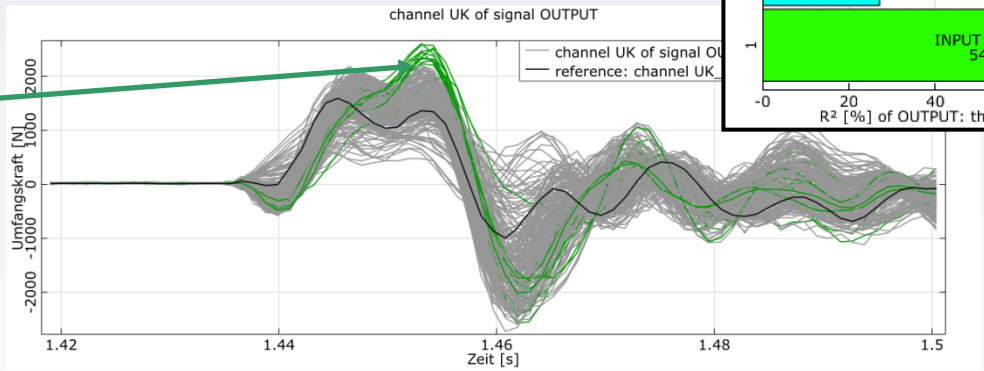
1. Scanning the Design Space with optimized LHS, variation and correlation are investigated



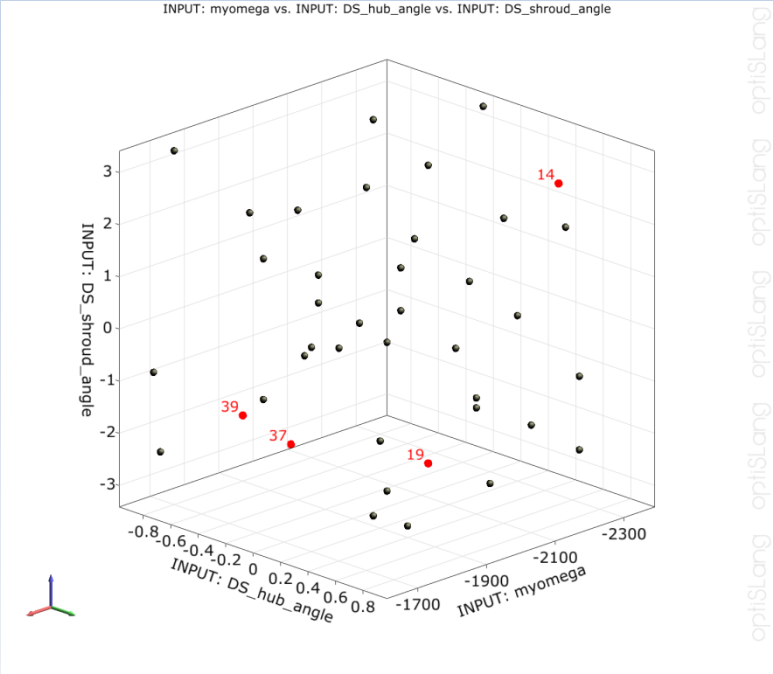
2. Identification of important variables
   - Check Variation of Design Space
   - Check Coefficient of Importance

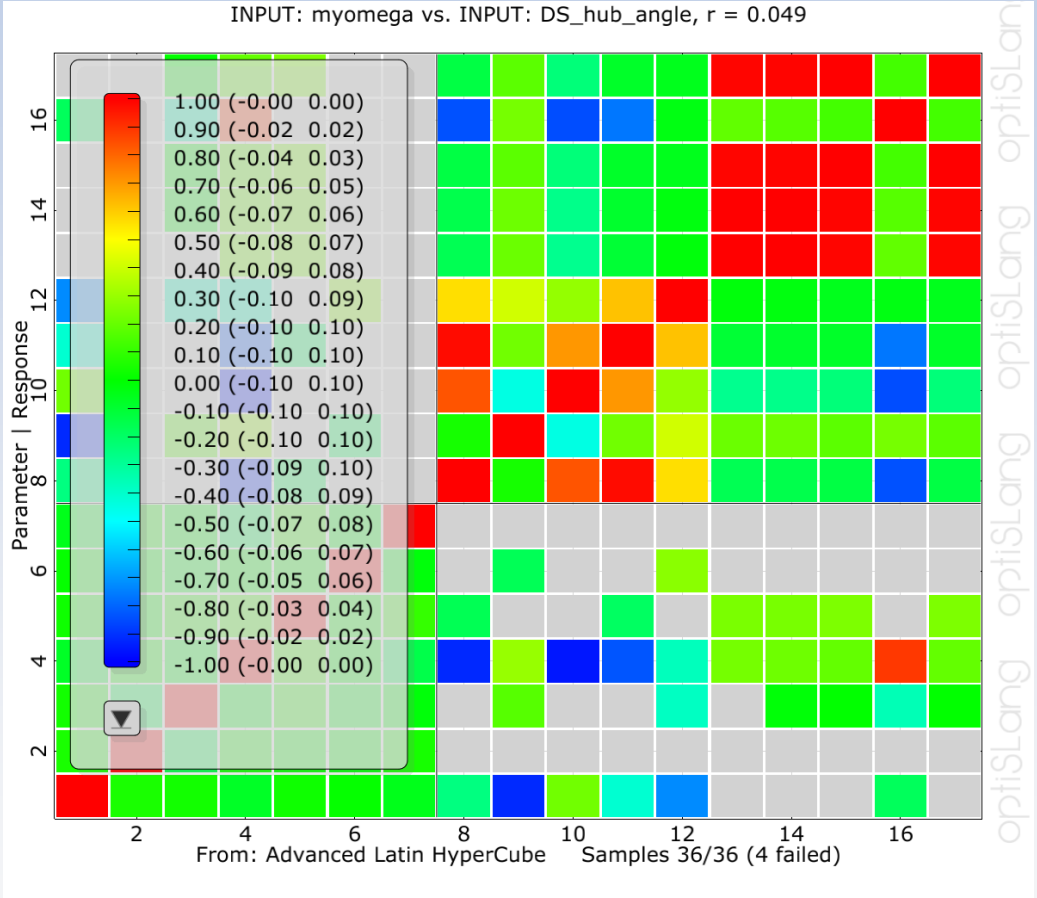






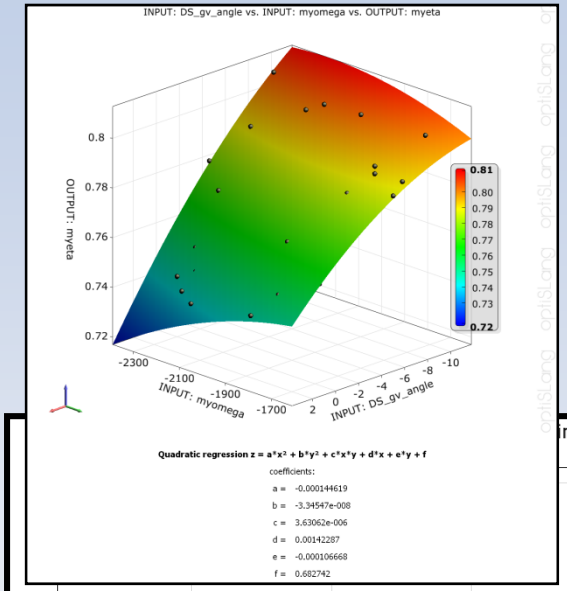
Latin Hypercube and Confidences

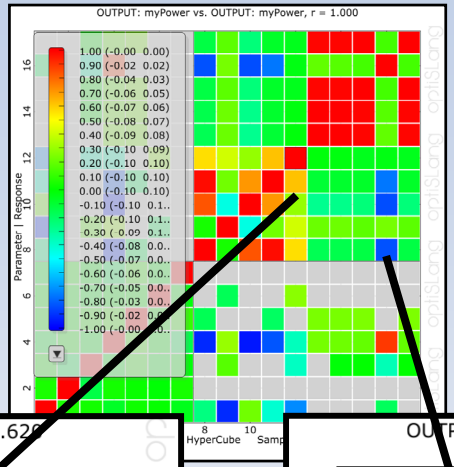


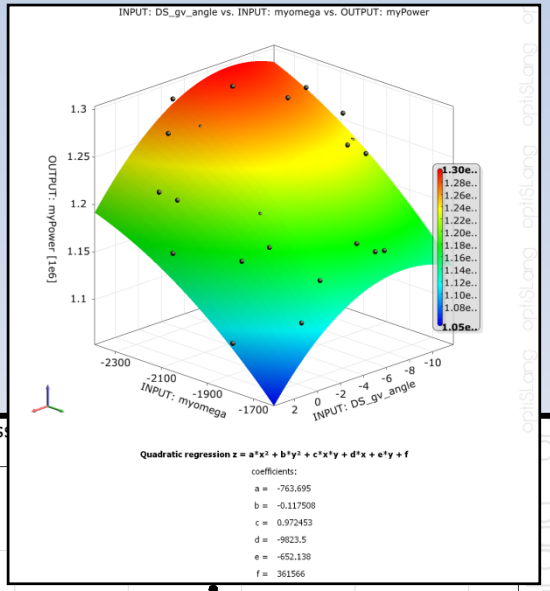


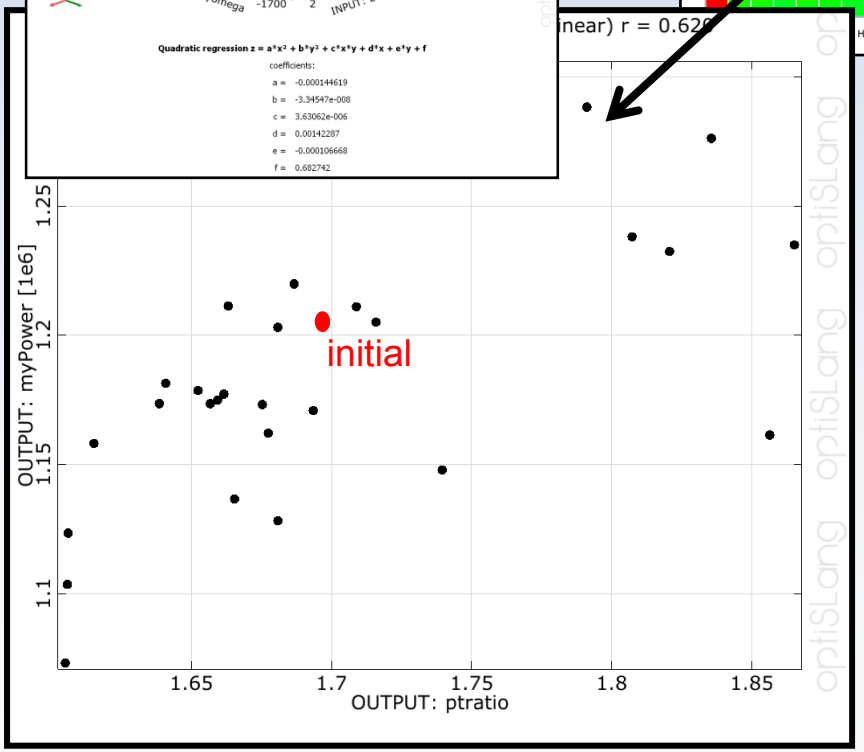
- n = 7 design variables
- N = 40 design evaluations (4 failed)
- Confidence levels are quite acceptable

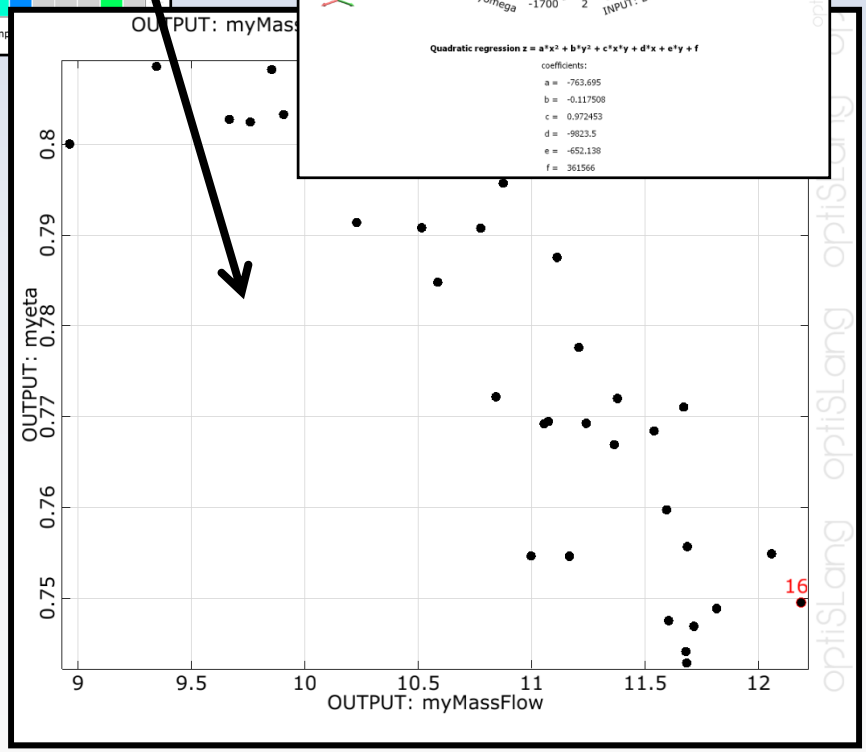
Variation of Design Space







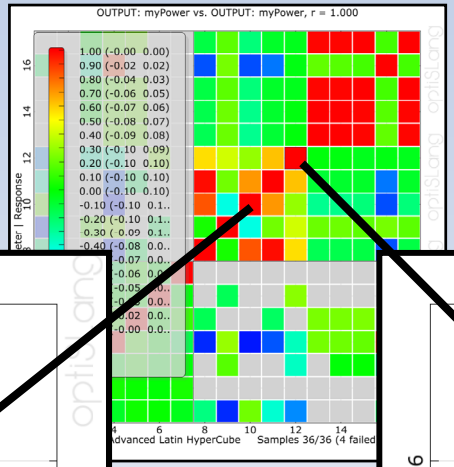


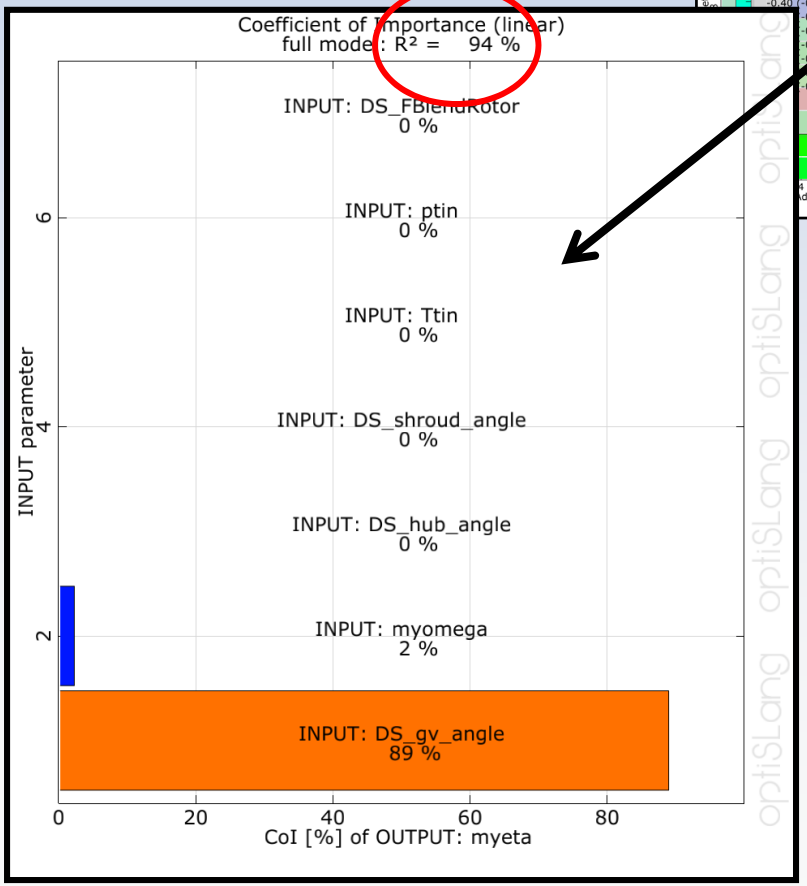


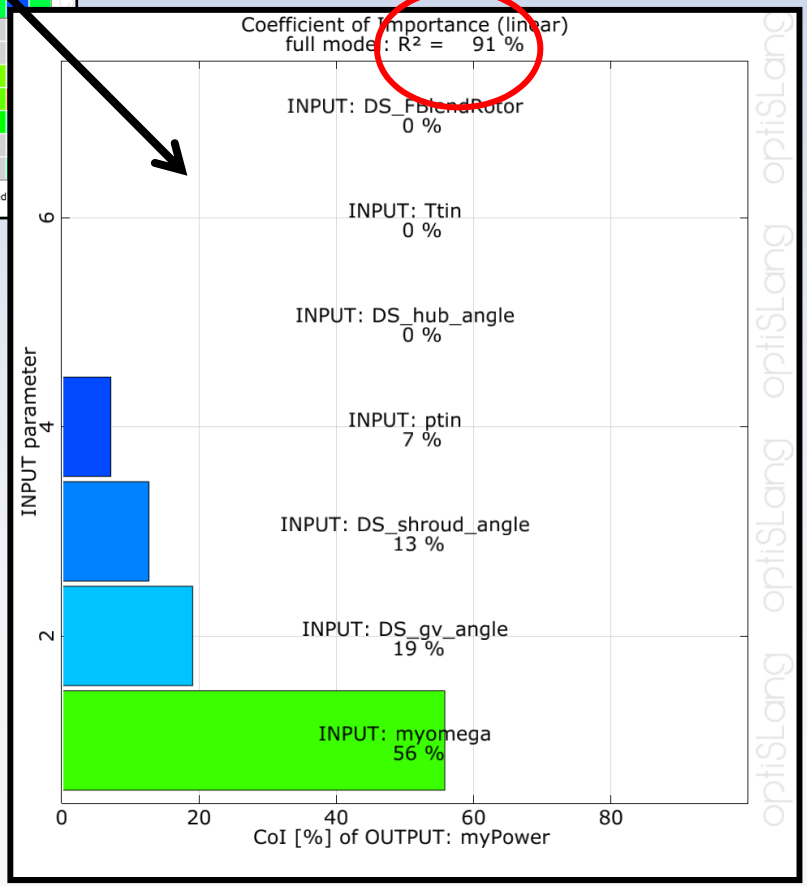
Coefficient of Importance, CoI

CoI=94%>80%

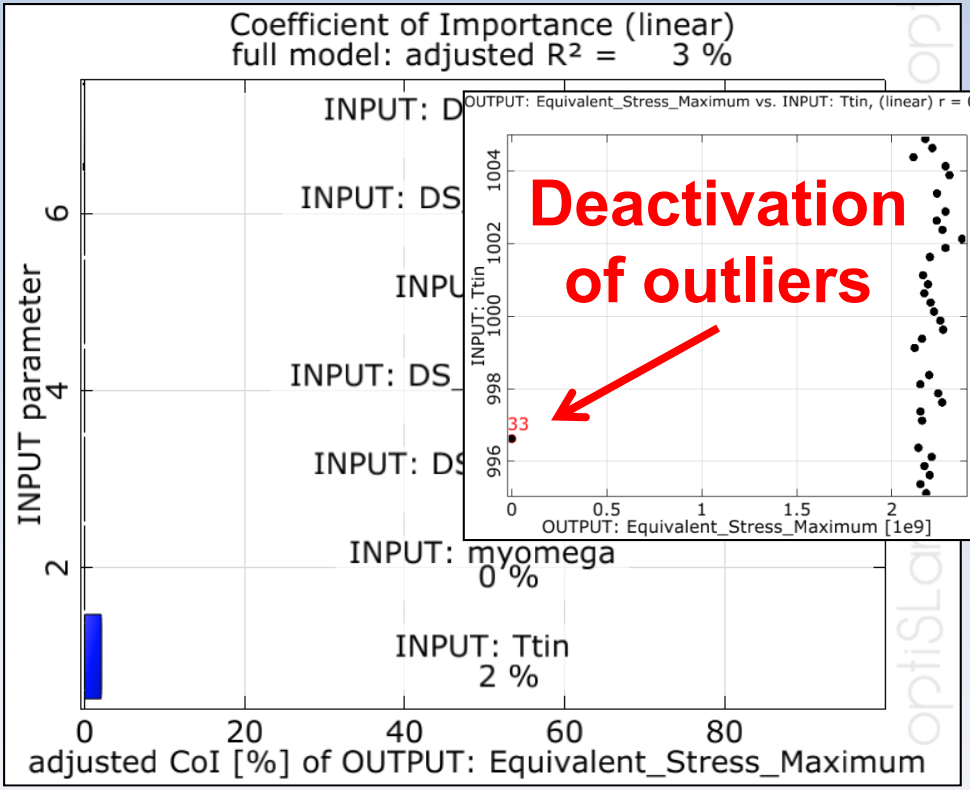
CoI=91%>80%

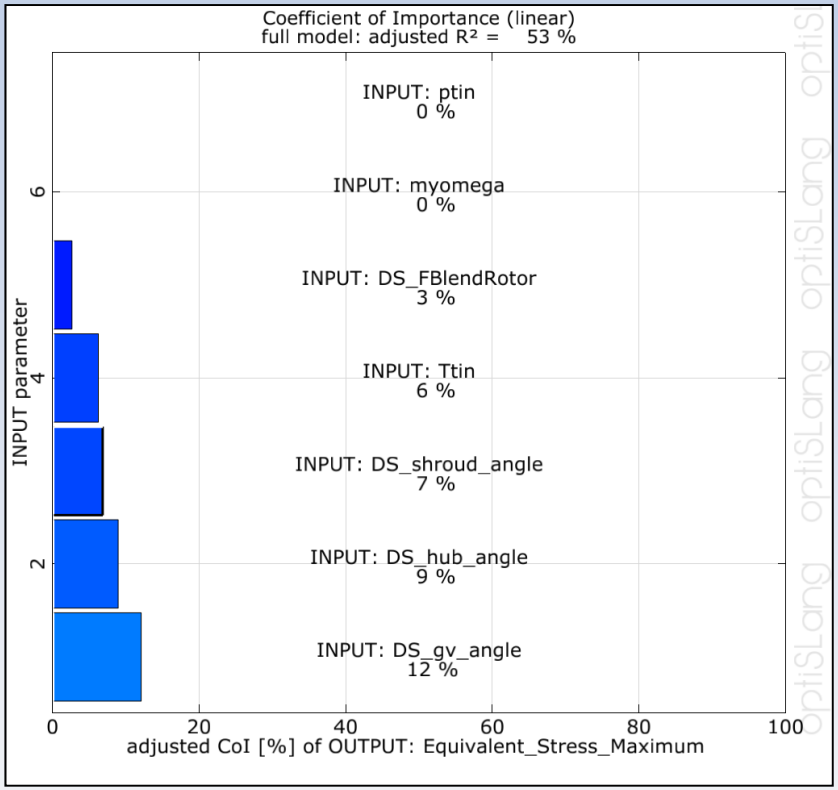






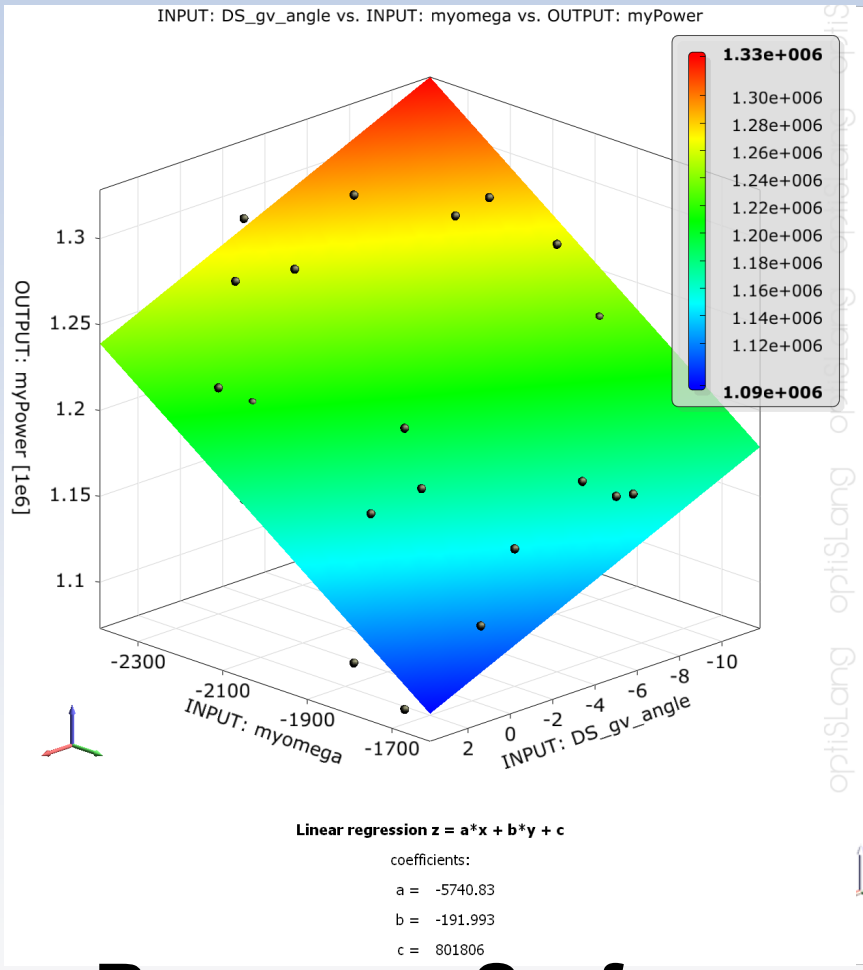
Coefficient of Importance, CoI

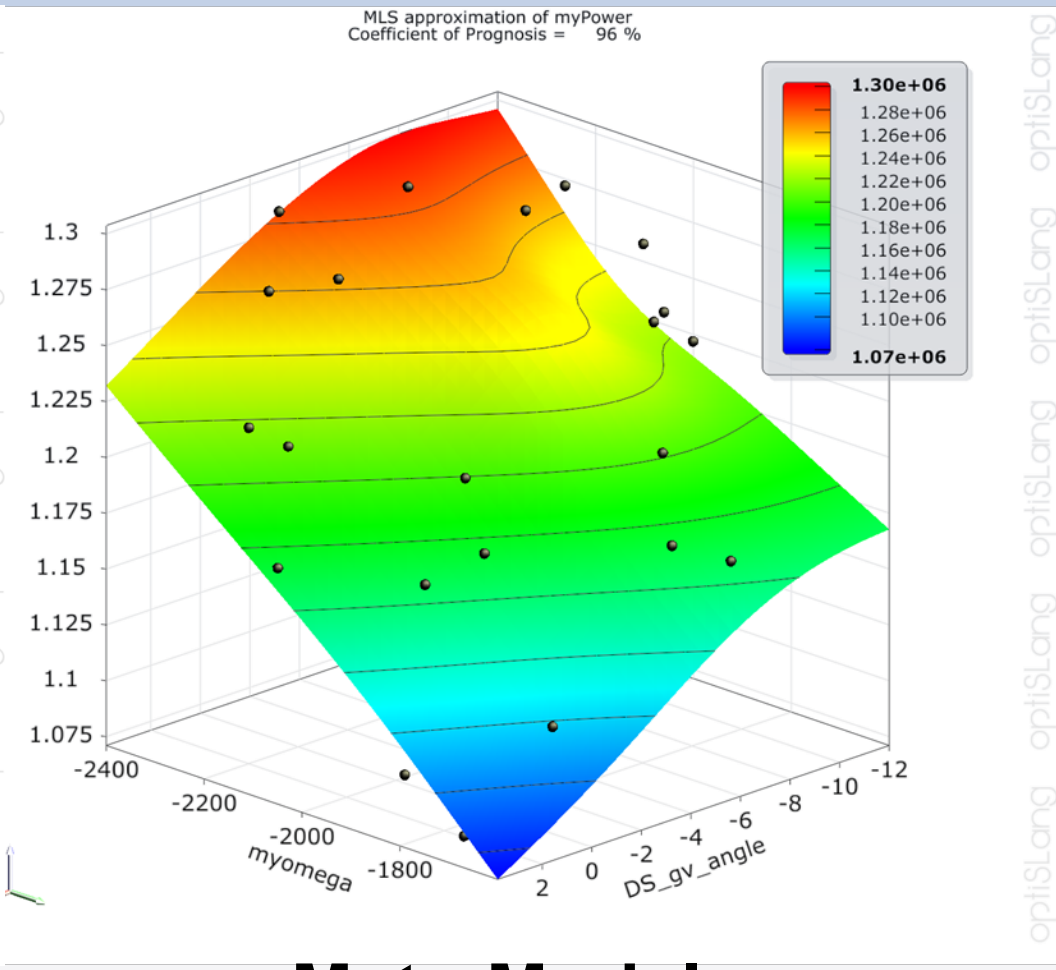




50% variance of the stress variation can be explained by the given n = 7 design variables

Meta-Model of Prognosis, MoP

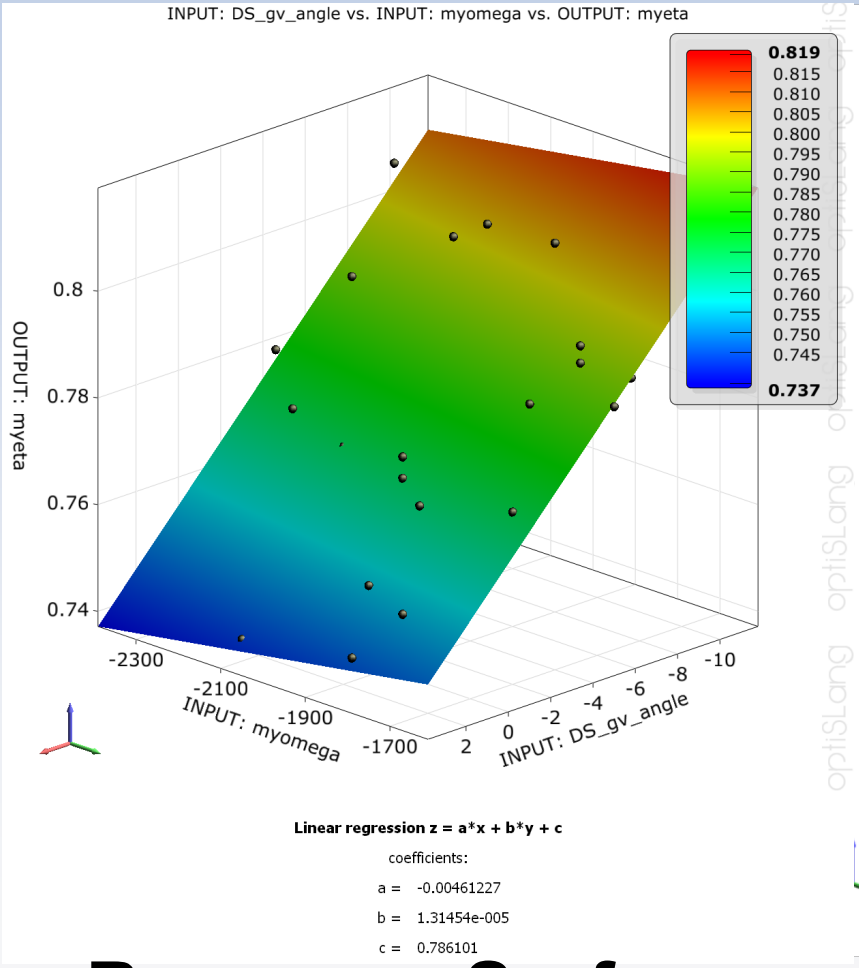


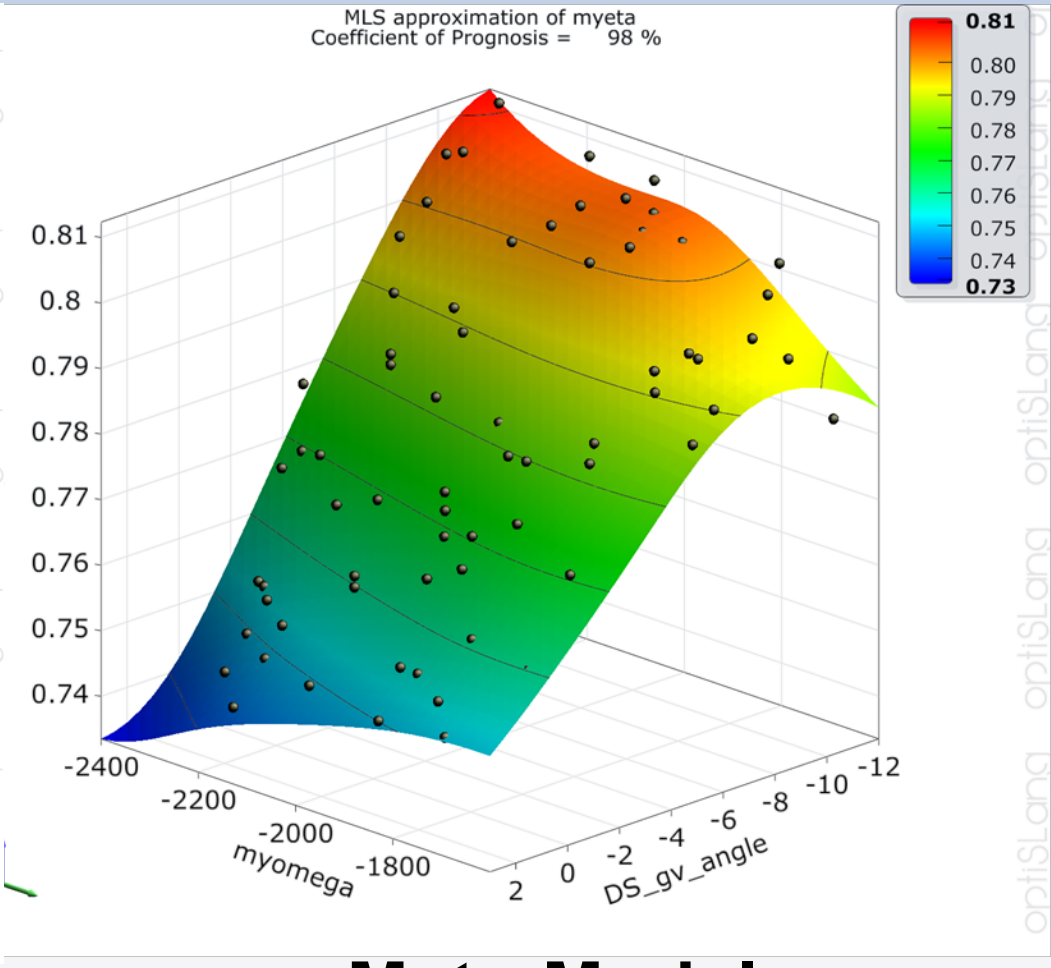


Response Surface Output: Power

Meta-Model Output: Power

Meta-Model of Prognosis, MoP

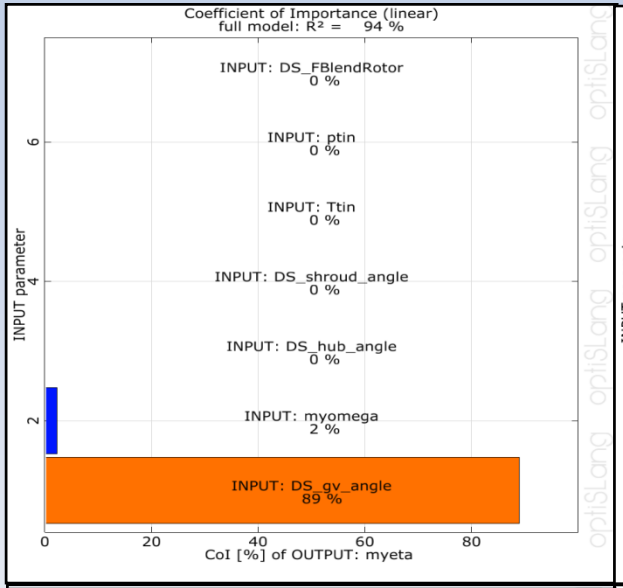


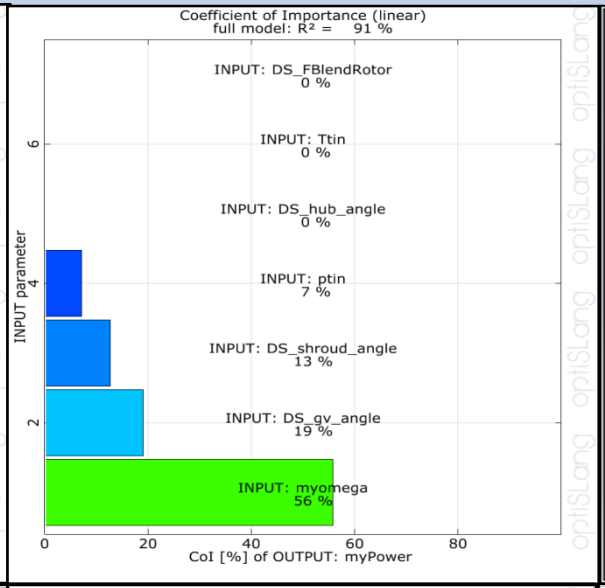


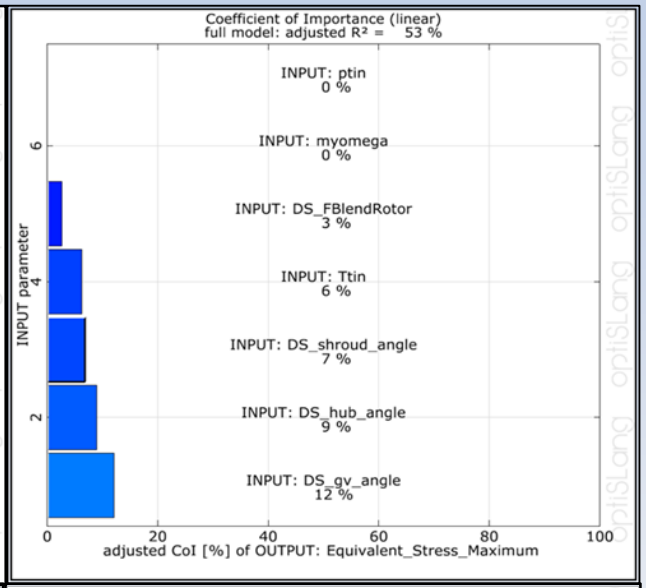
Response Surface
Output: Efficiency

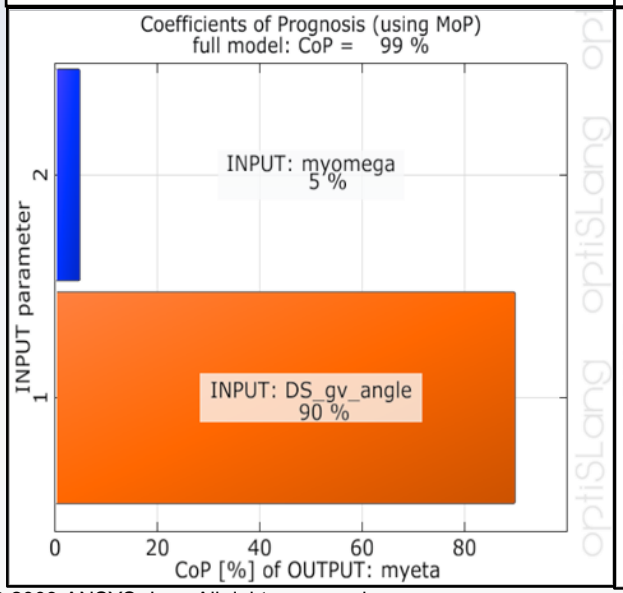
Meta-Model
Output: Efficiency

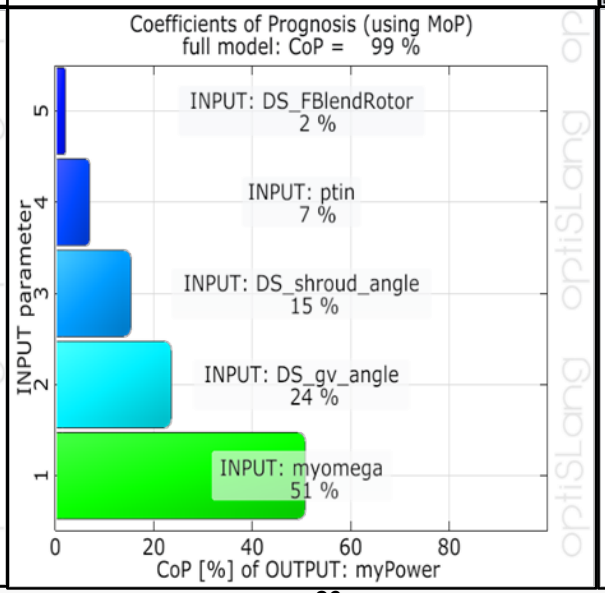
Coefficient of Importance, CoI vs. Coefficient of Prognosis, CoP

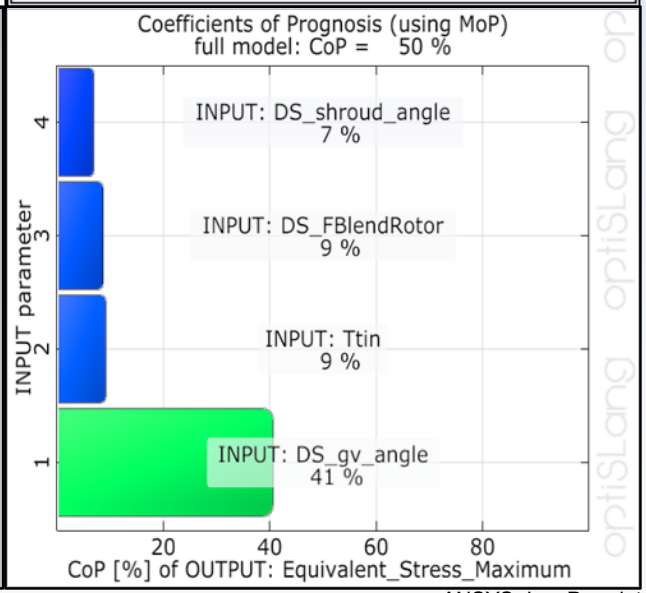




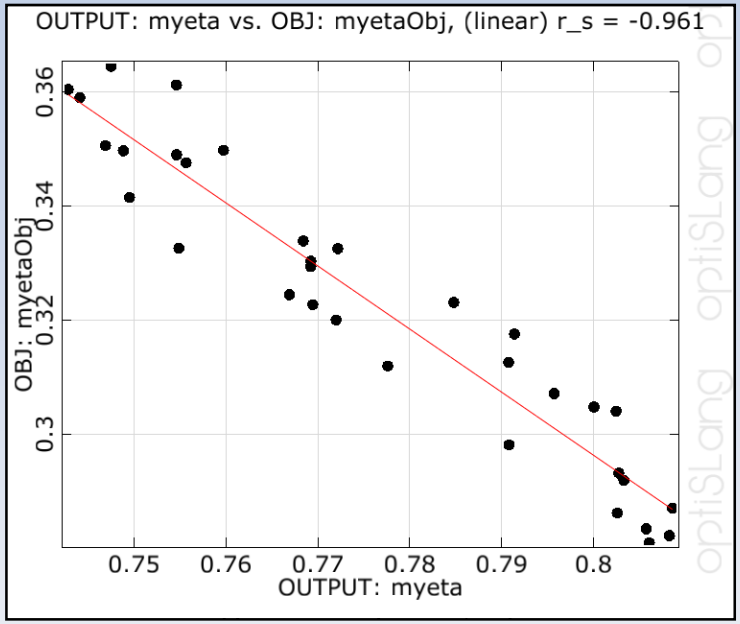


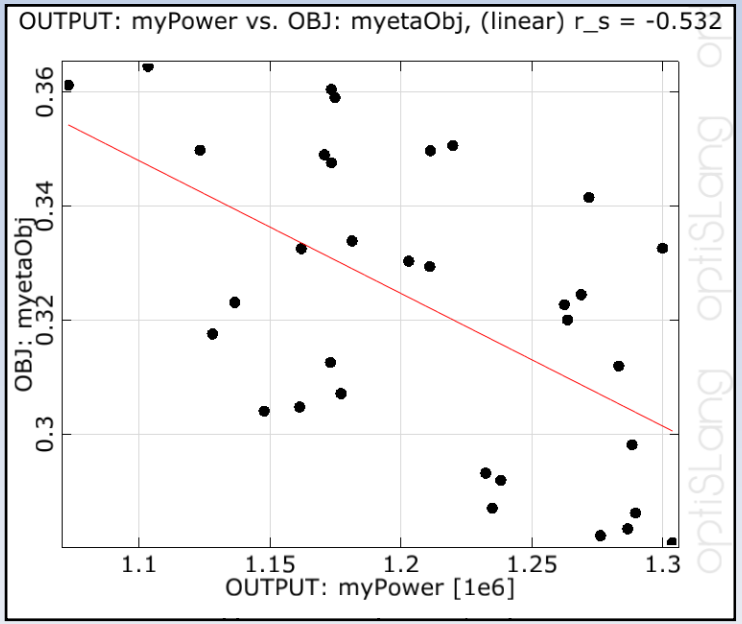






Objective Function

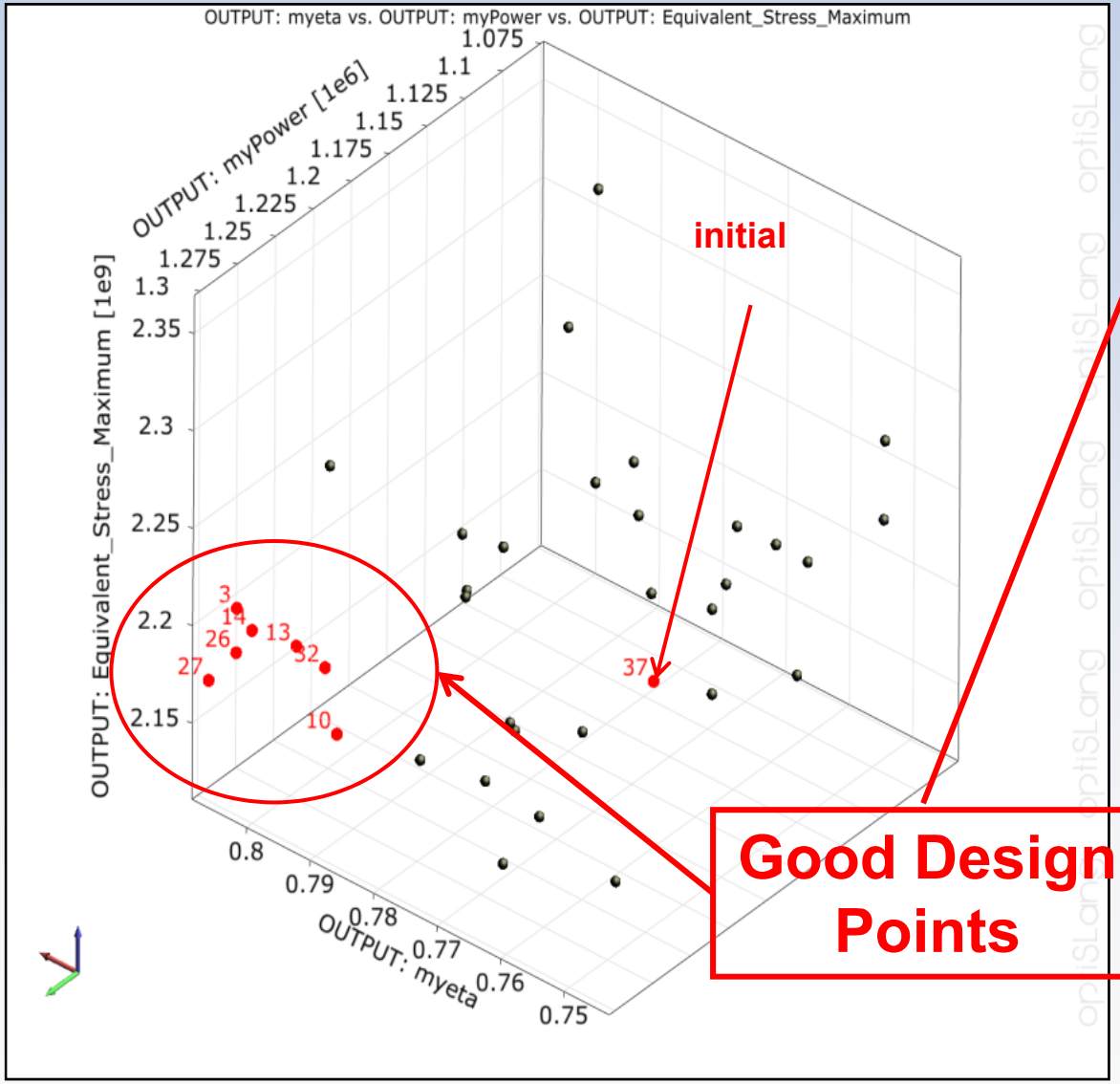


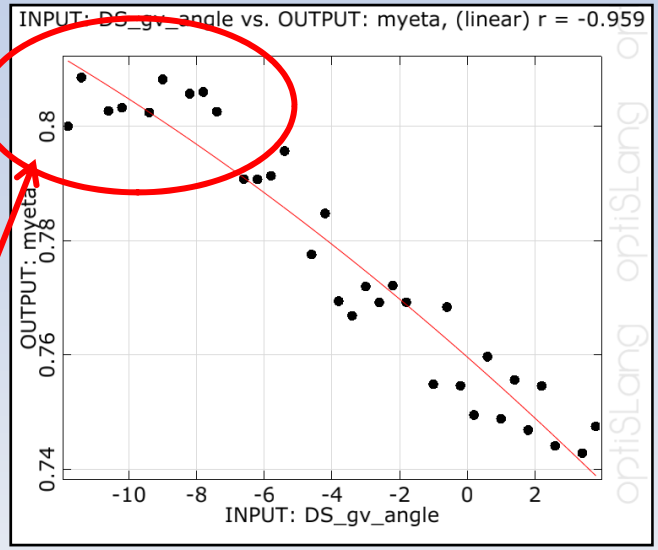


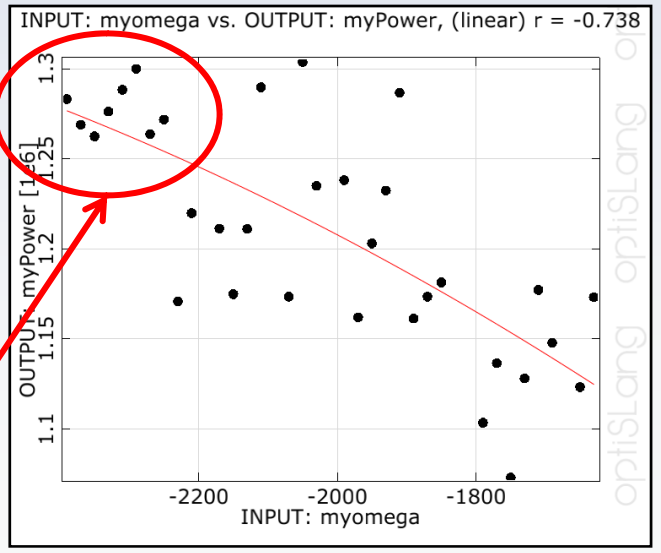
Target Function for Optimization:

$$f_{\text{Target}} = (1-\eta) + \frac{1}{4} \cdot \left(1 - \frac{P}{2[MW]}\right)^{!} = \min; \quad \sigma_{seqv} < \sigma_{Limit}$$

Anthill Plots Objective

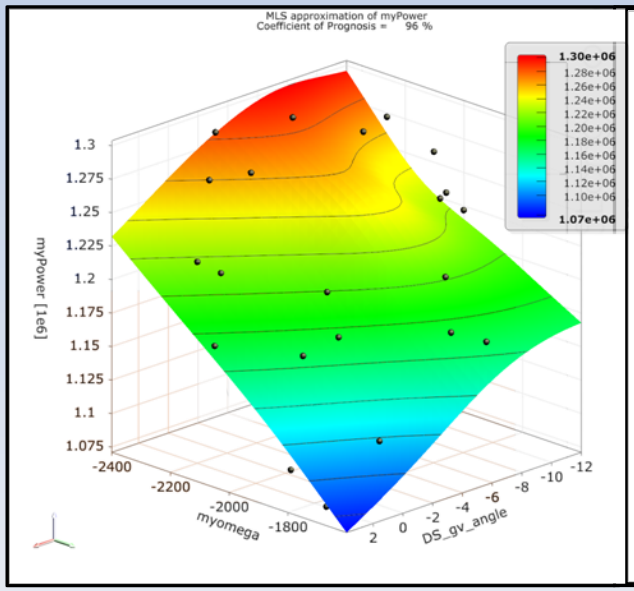


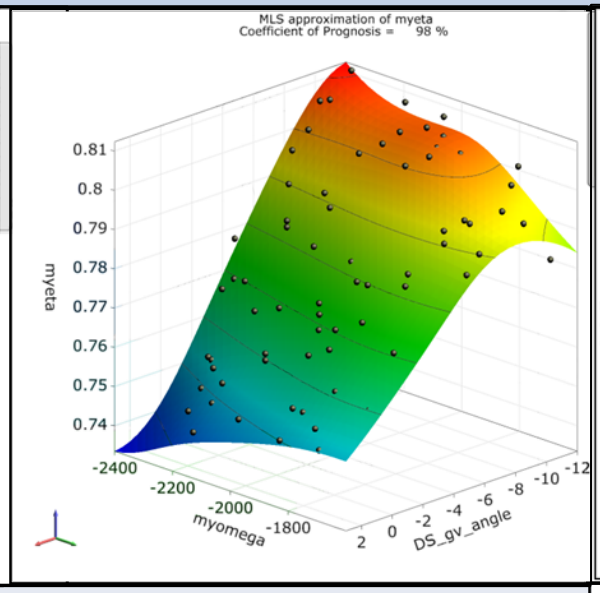


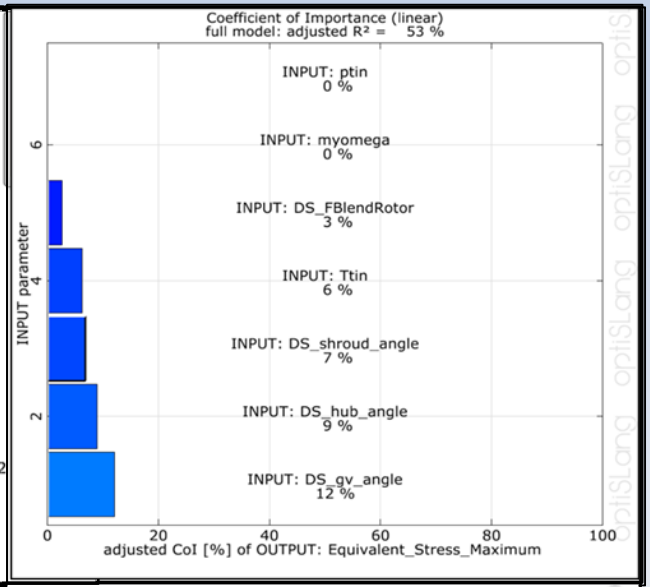


Good Design Points

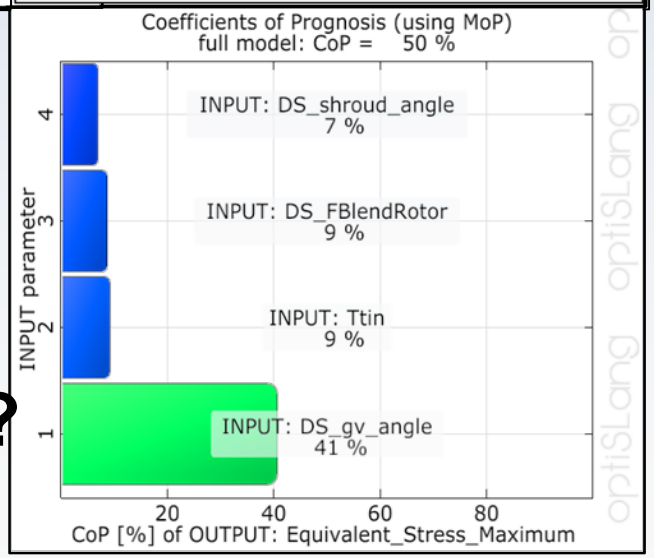
Conclusion Sensitivity Analysis

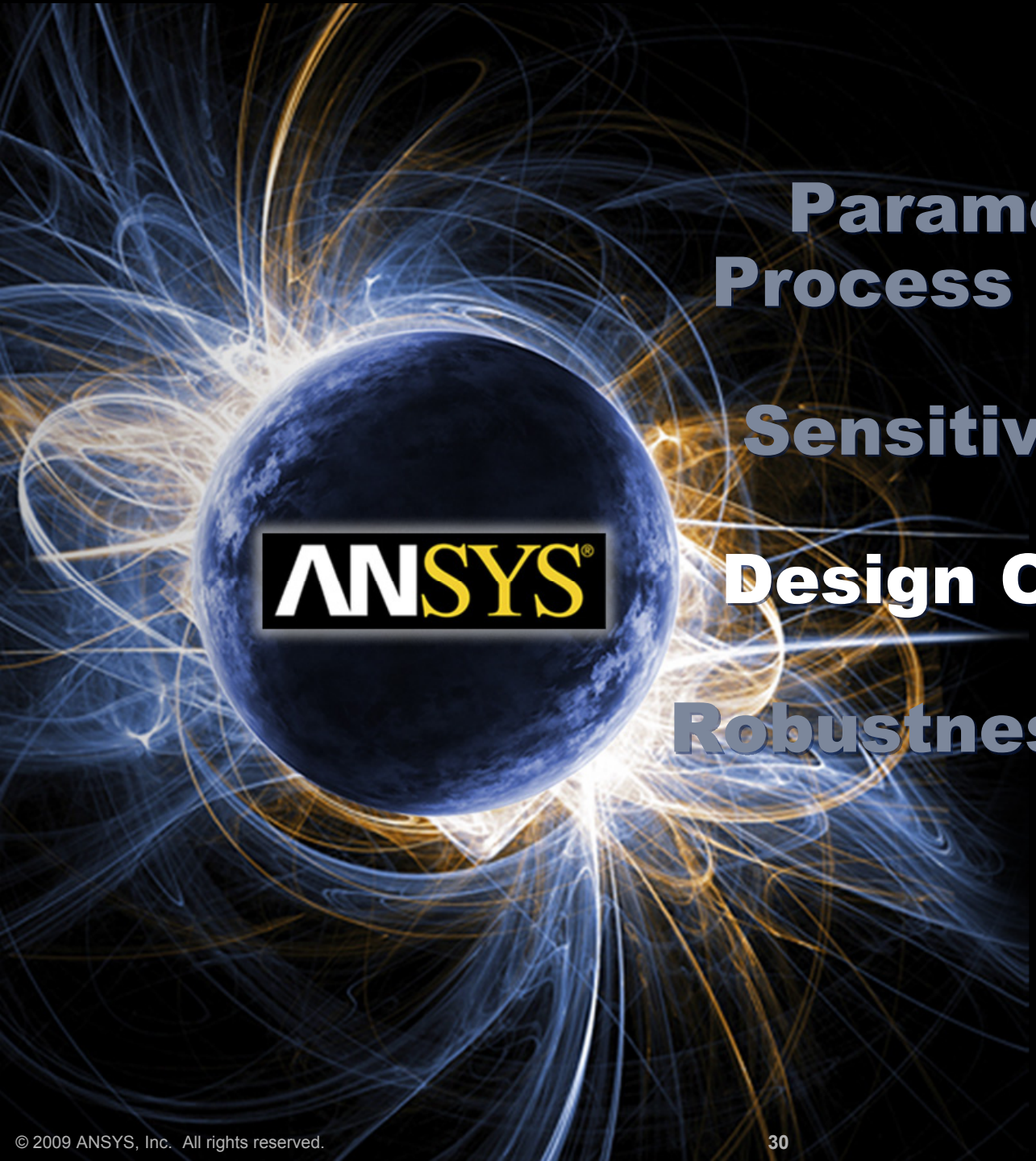






- Is Sensitivity reliable, CoI?
- Is Sensitivity reliable, CoP?
- Is Sensitivity plausible, physics?





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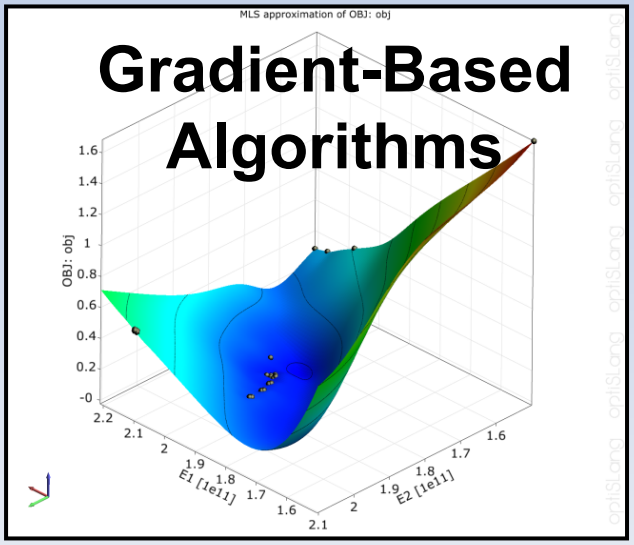
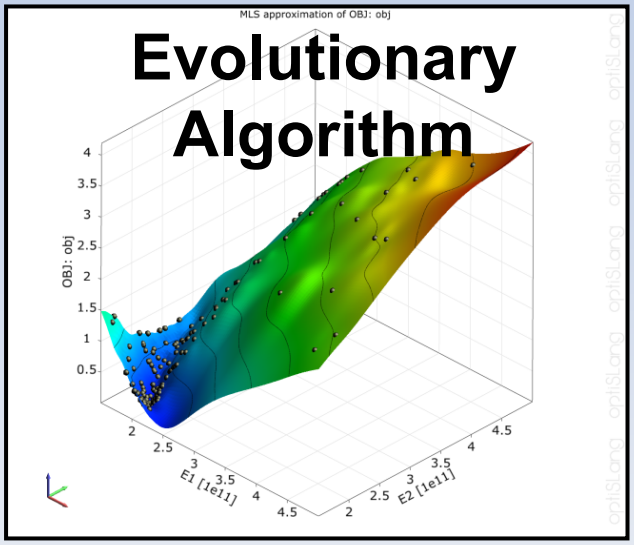
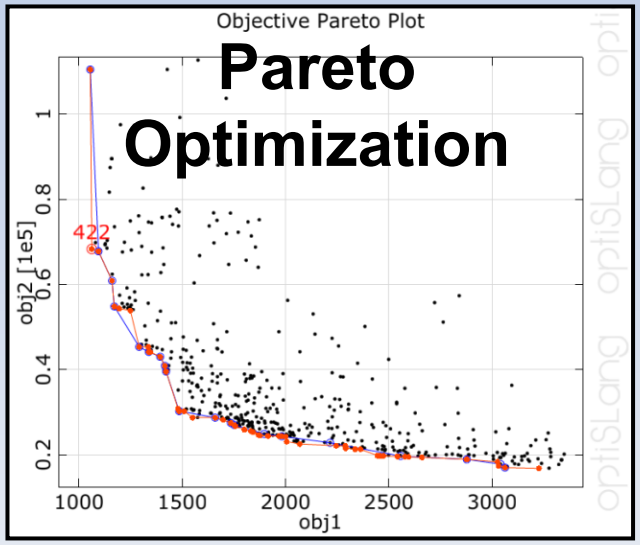
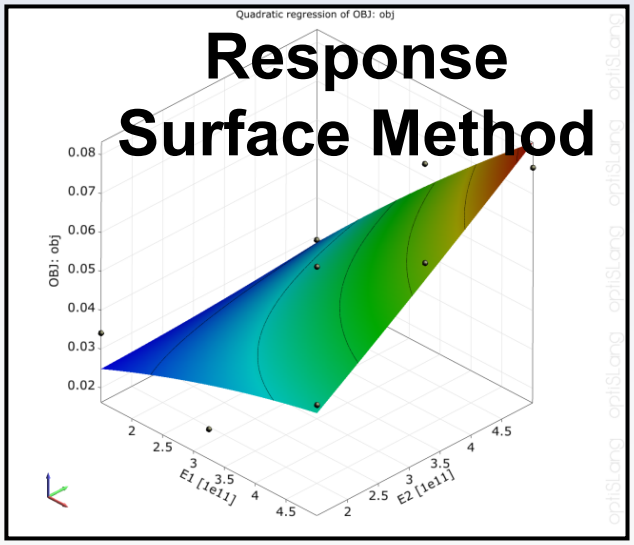
Robustness Evaluation

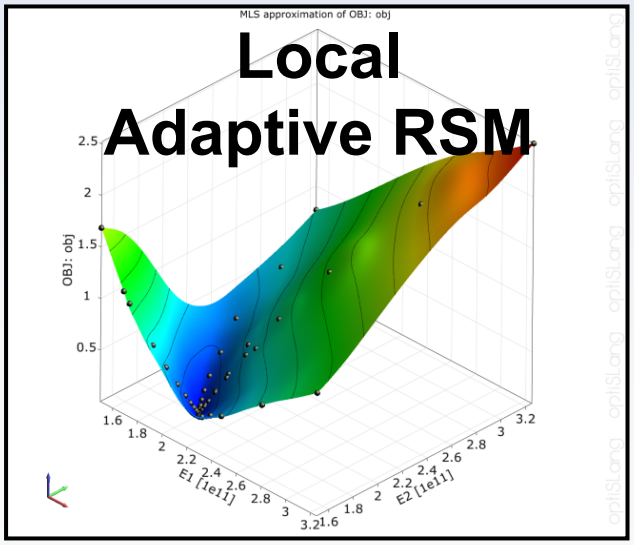
Design Optimization

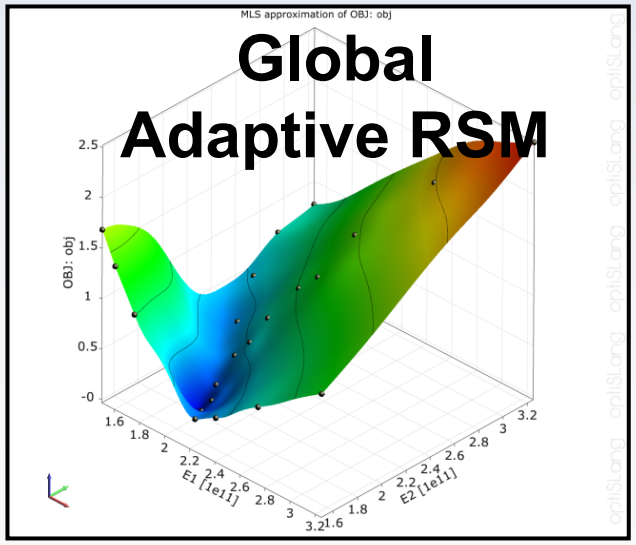












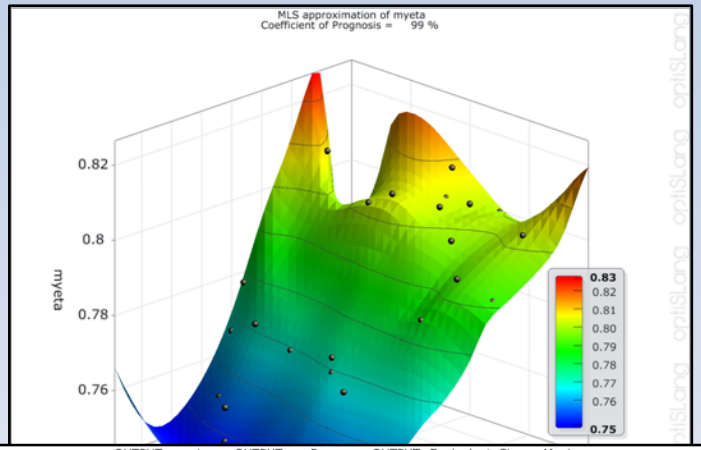
Optimization Strategy

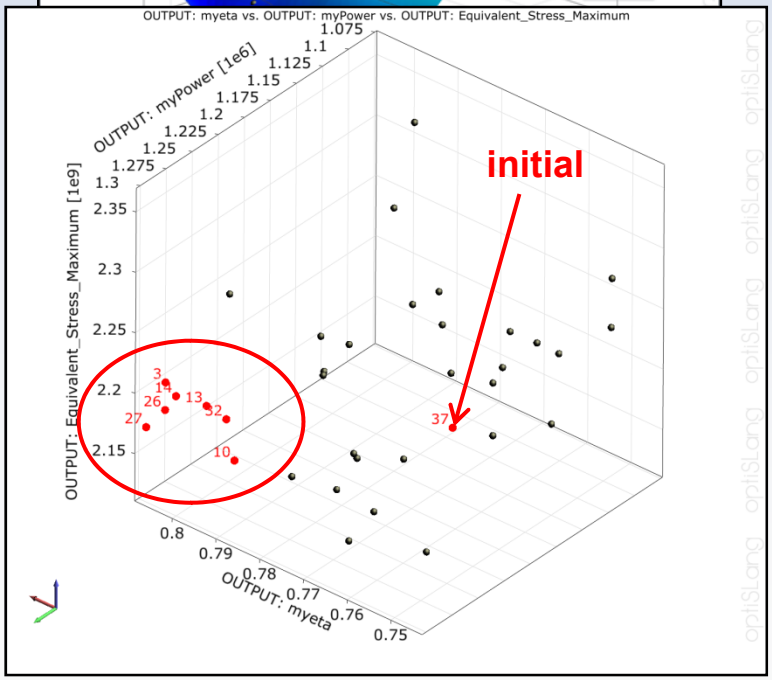
Sensitivity Analysis

- Shows Potential
- Indicates multiple local optima
- No parameter reduction

Strategy:

- Global search, EA
- Start design(s) from SA
- Local improvement, ARSM

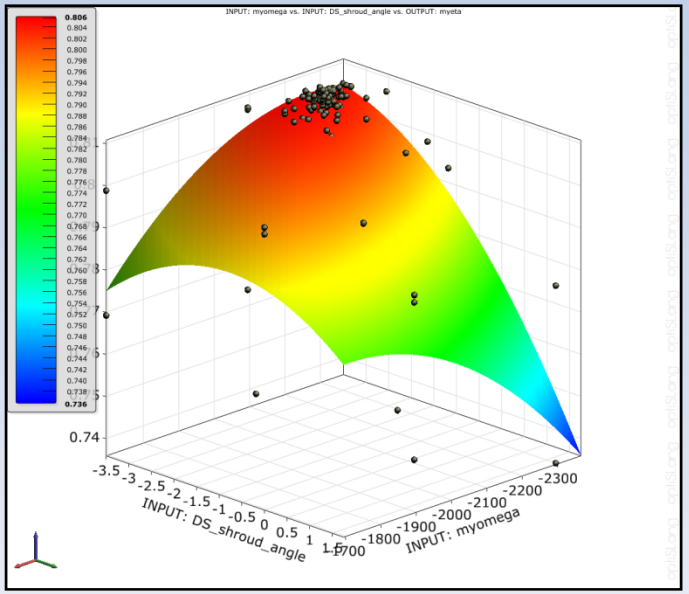




Outlook Strategy

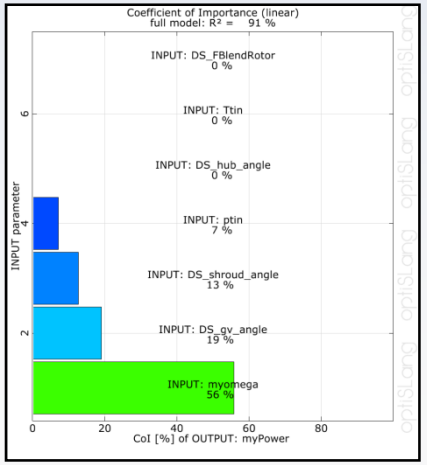
Sensitivity Analysis

- Shows Potential
- Indicates global optimum
- Parameter reduction



Strategy:

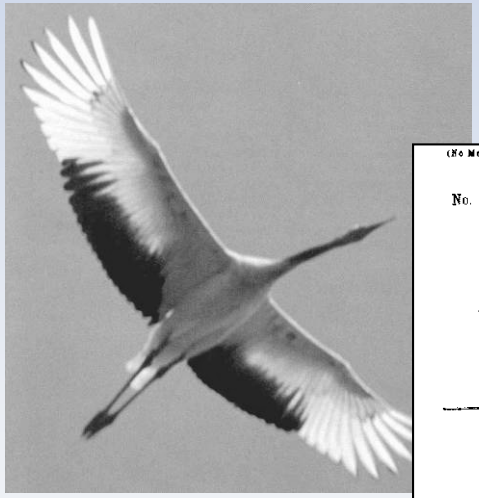
- Pre-optimization in sub space, ARSM
- Local improvement, EA (full space)
- Start design(s) from ARSM

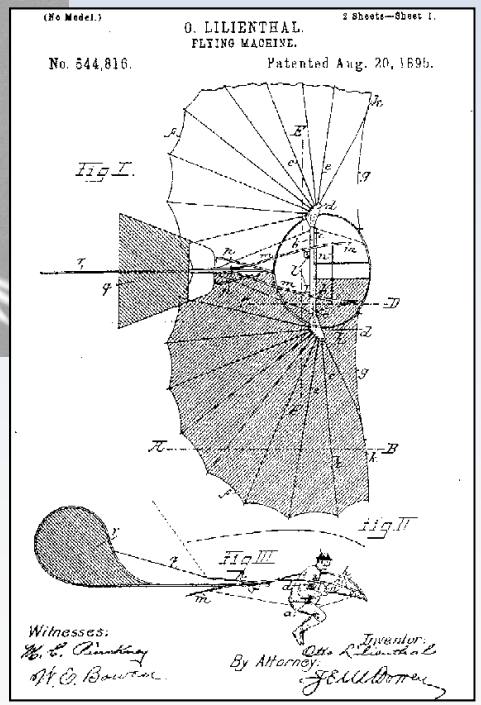


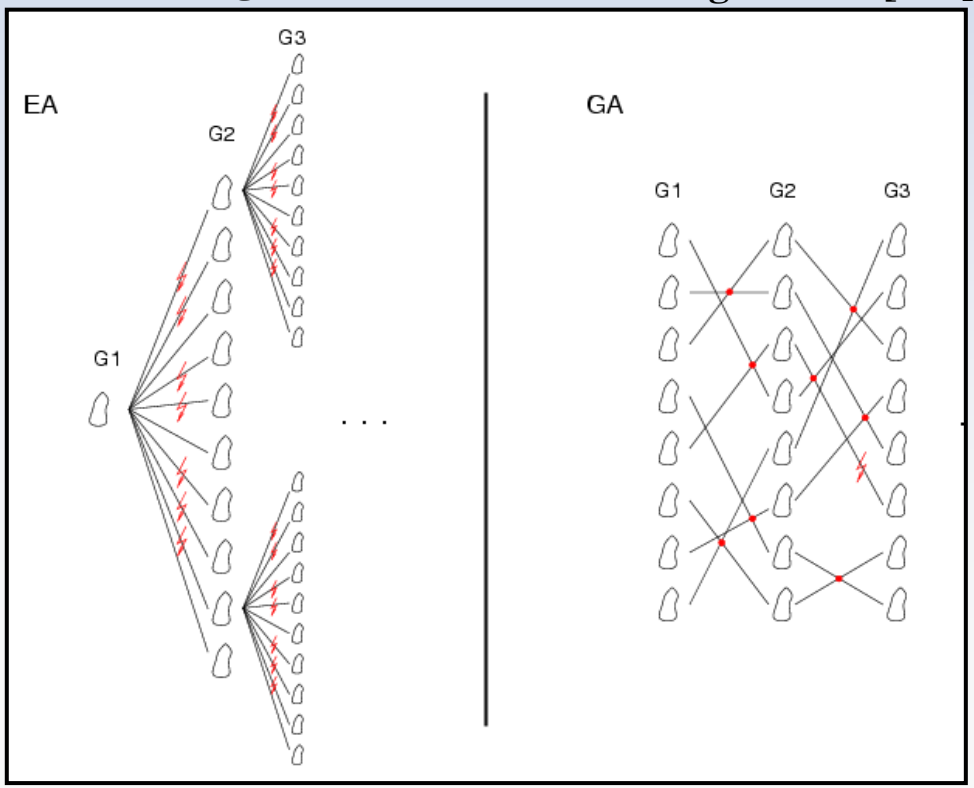
Evolutionary Algorithms (EA)

Optimization in Nature:

- Survival of the fittest
- Evolution due to mutation, recombination and selection

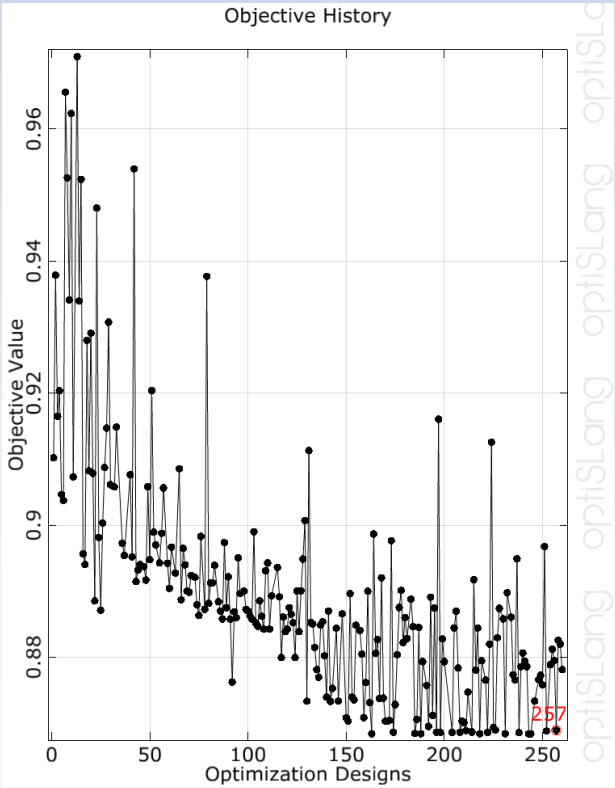


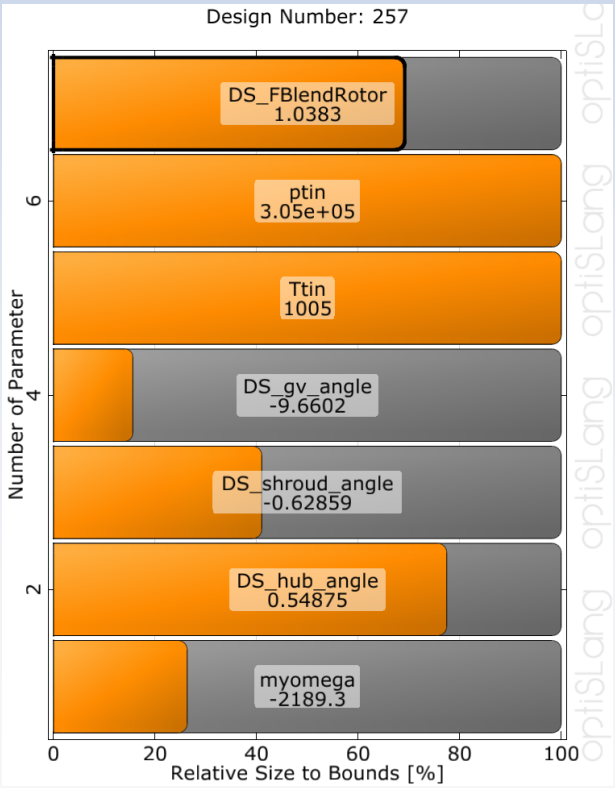


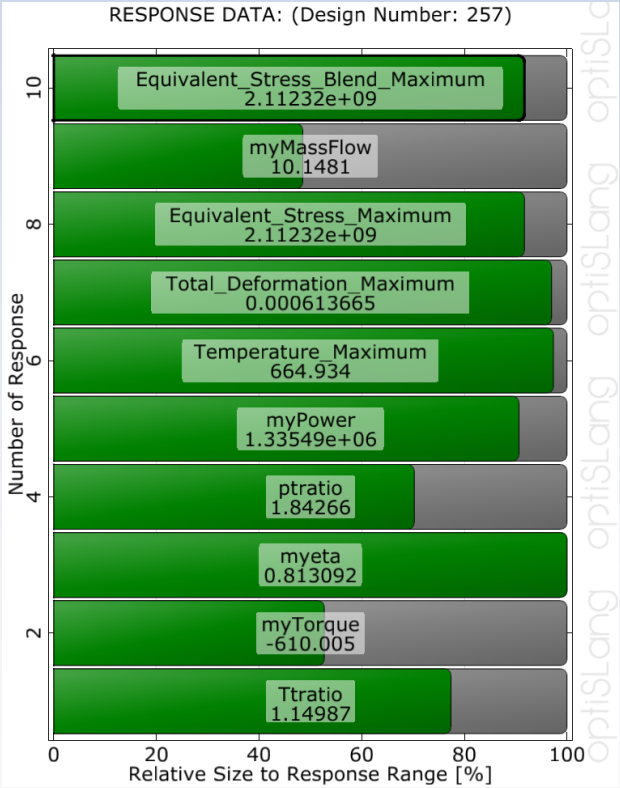


Evolutionary Algorithms (EA)

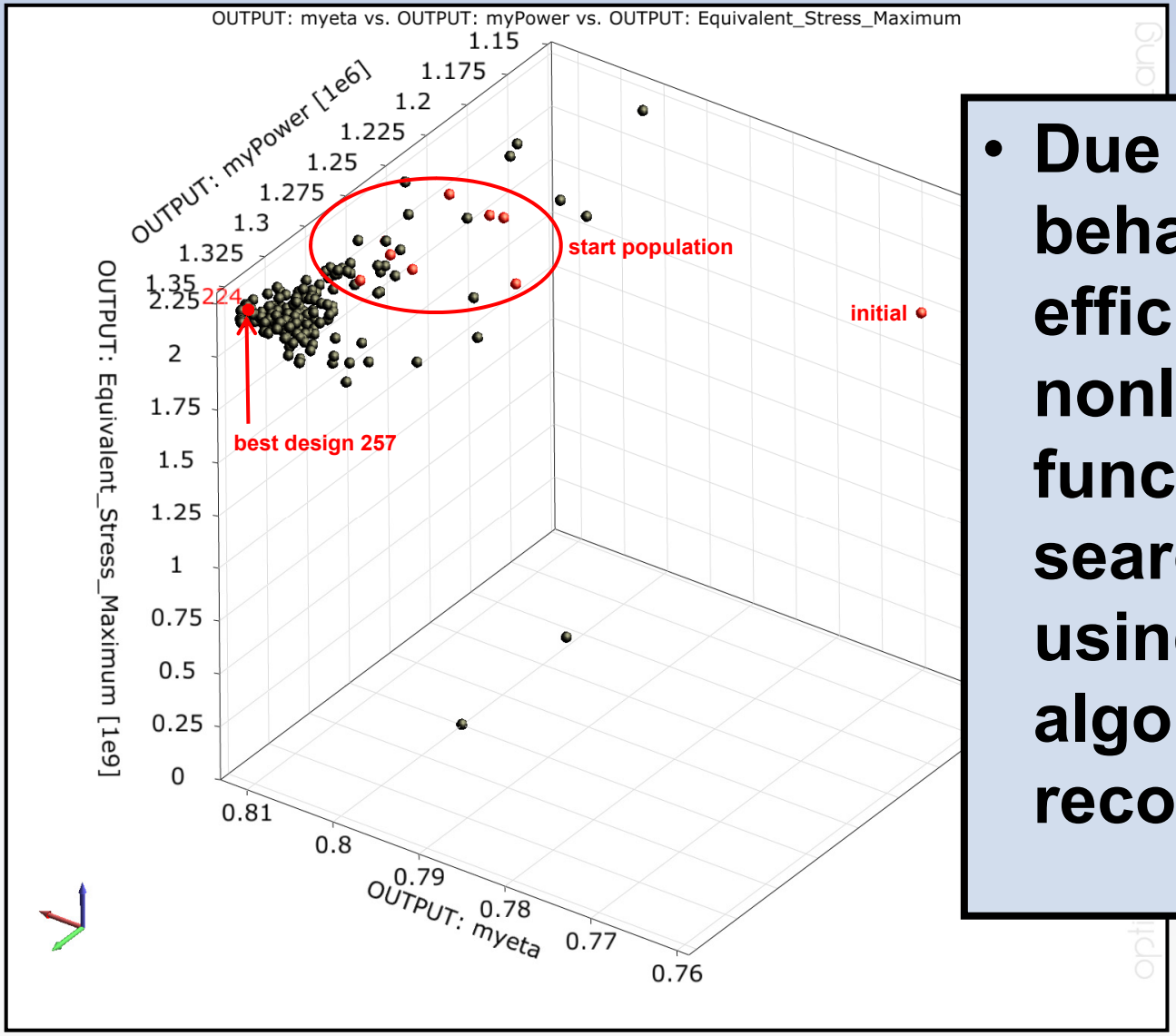
- **History of the Evolutionary Algorithm**





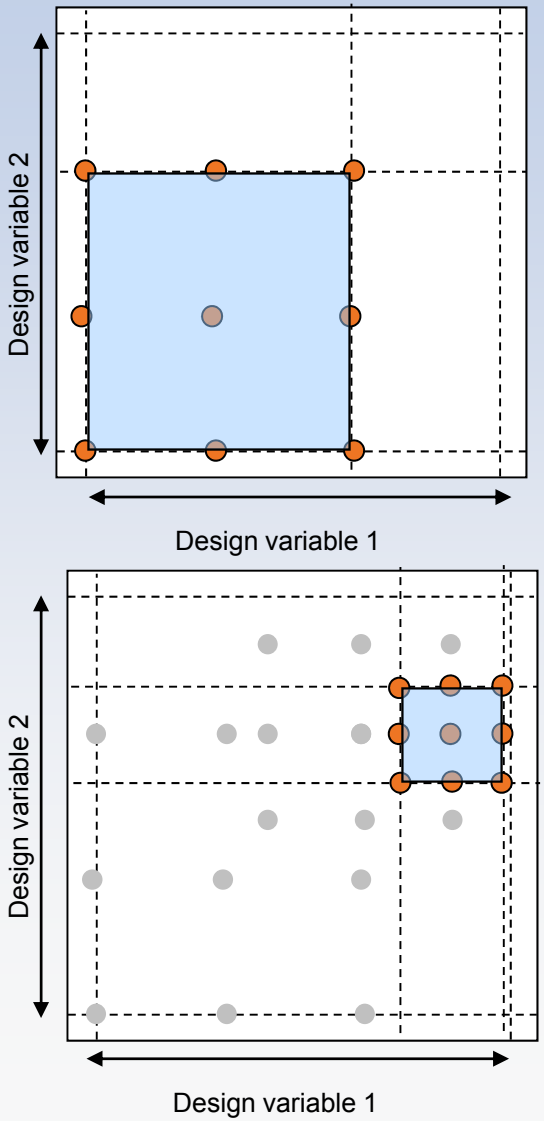


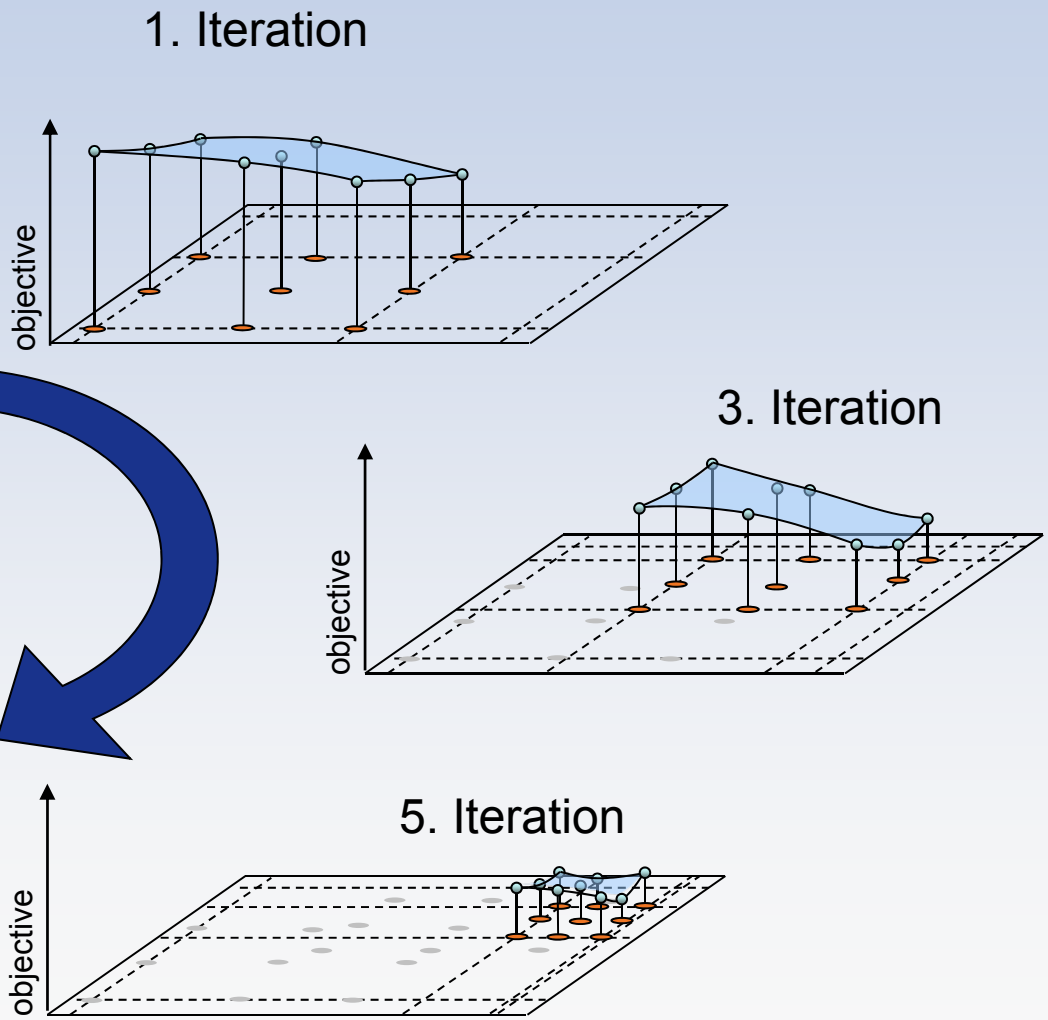
Evolutionary Algorithms (EA)



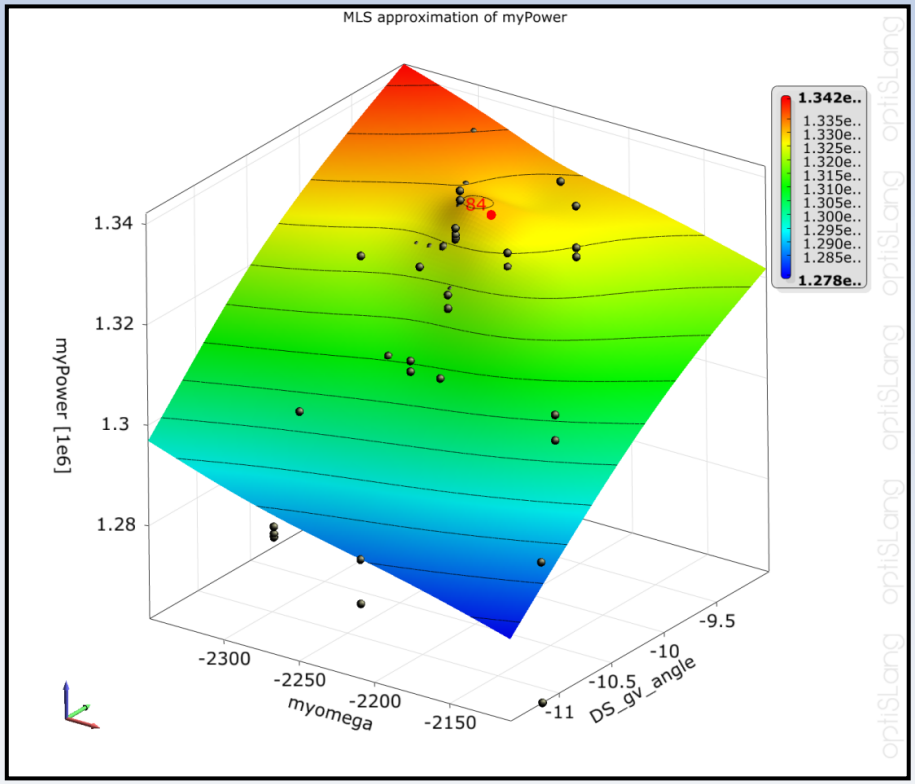
- Due the non-convex behavior of the efficiency and nonlinear power function a global search strategy using genetic algorithm is recommended

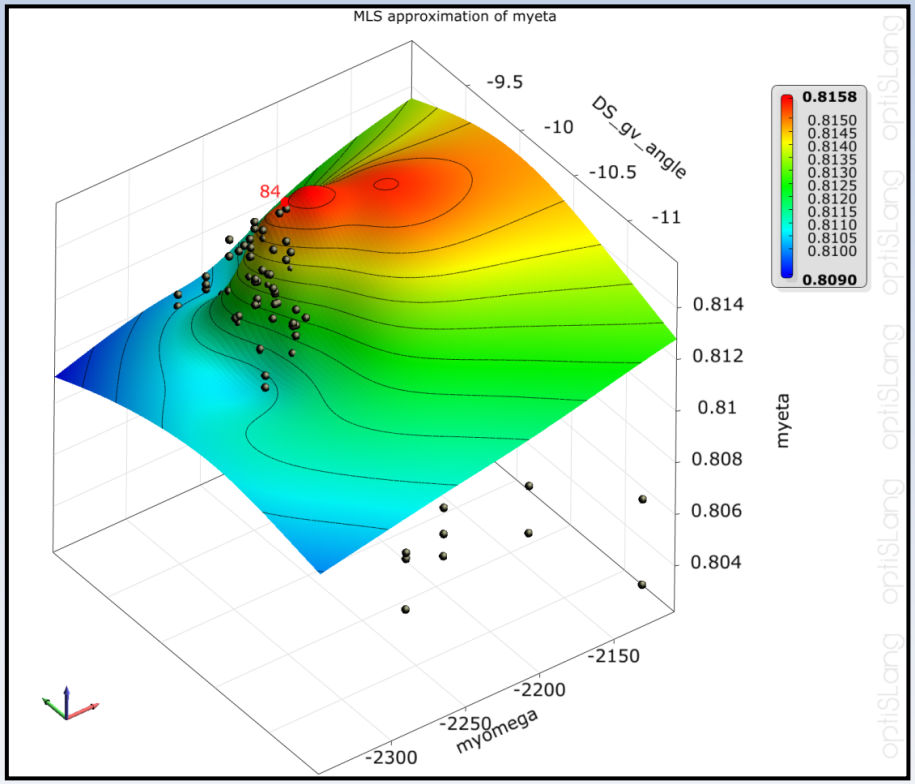
Adaptive Response Surface Methods (Local)





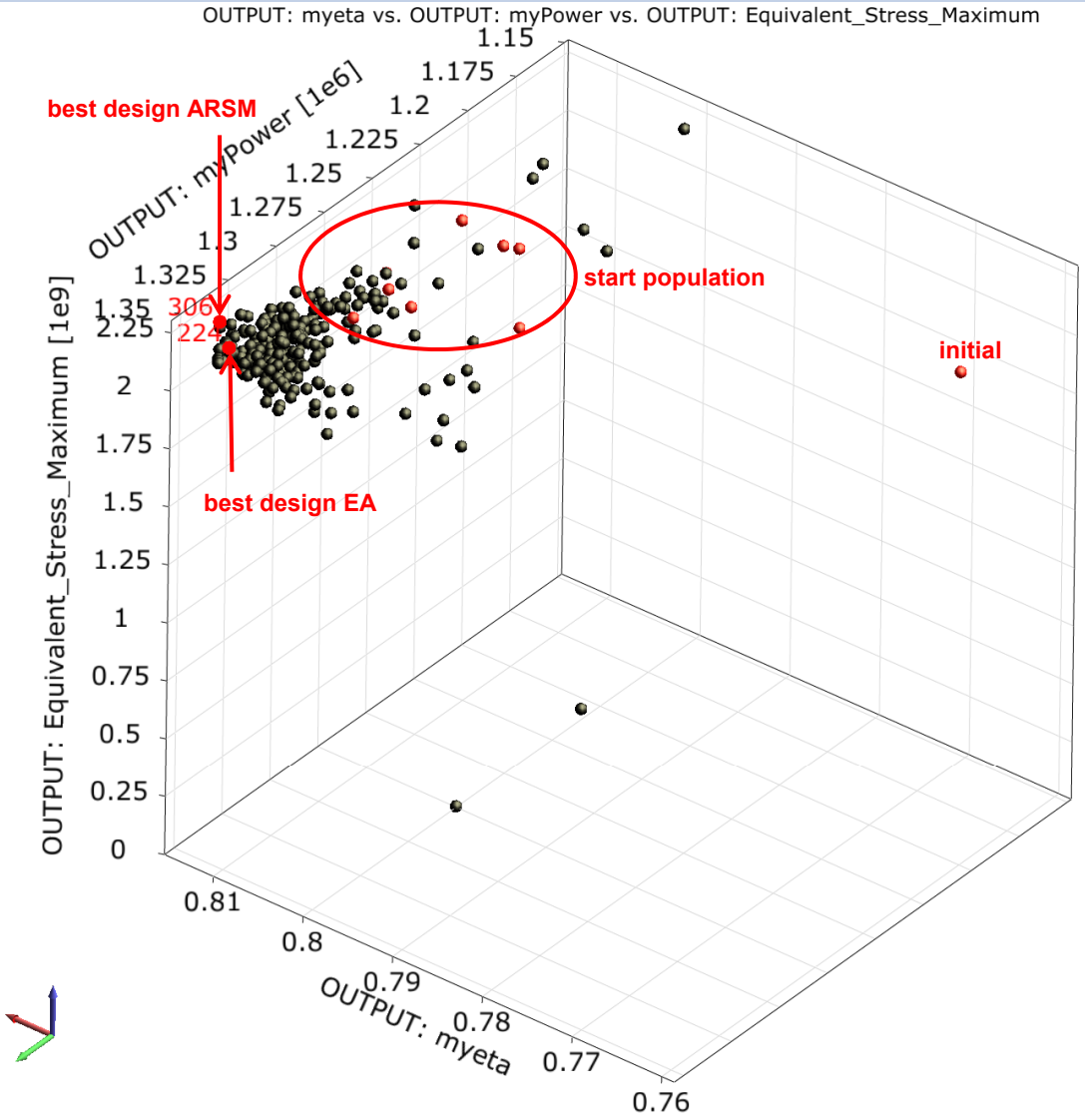
Adaptive Response Surface (ARSM)

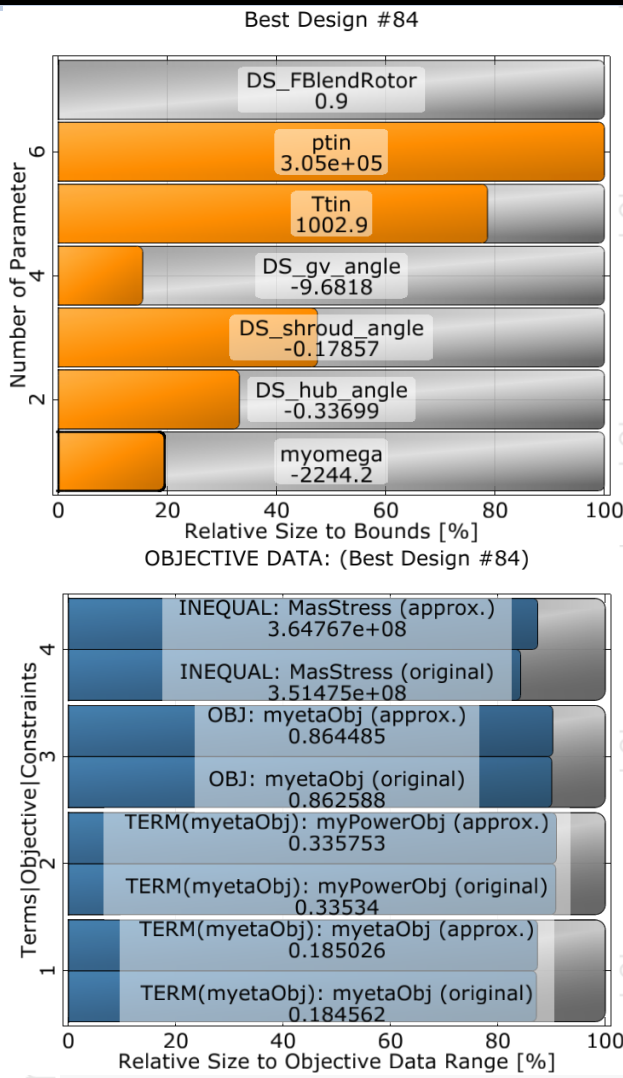




- The ARSM does not provide differentiable and smooth problems; very efficient for n < 15 design parameters
- Starting solution is based on the best design of the EA
- The design space is reduced to 20% around start solution

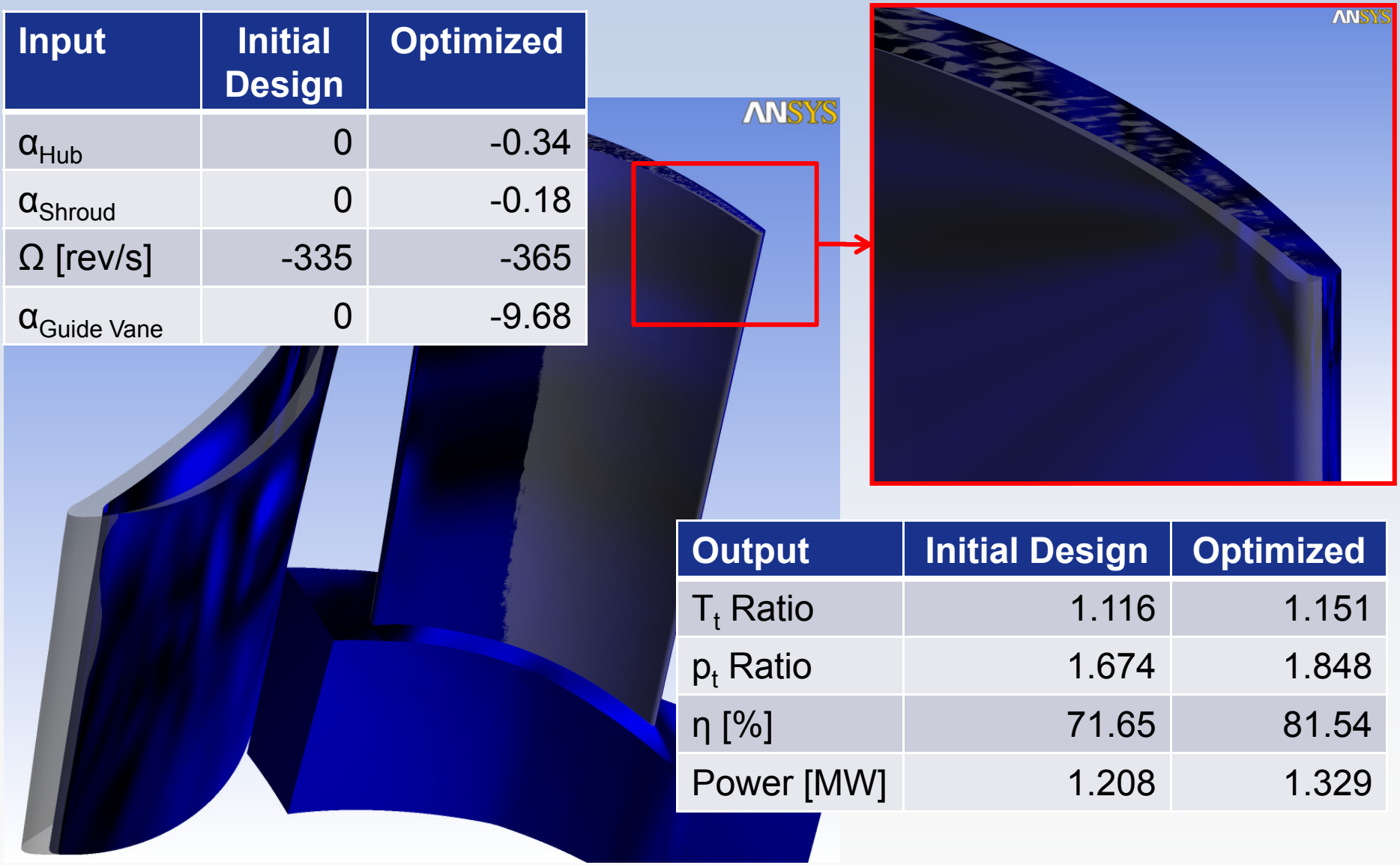
Adaptive Response Surface (ARSM)





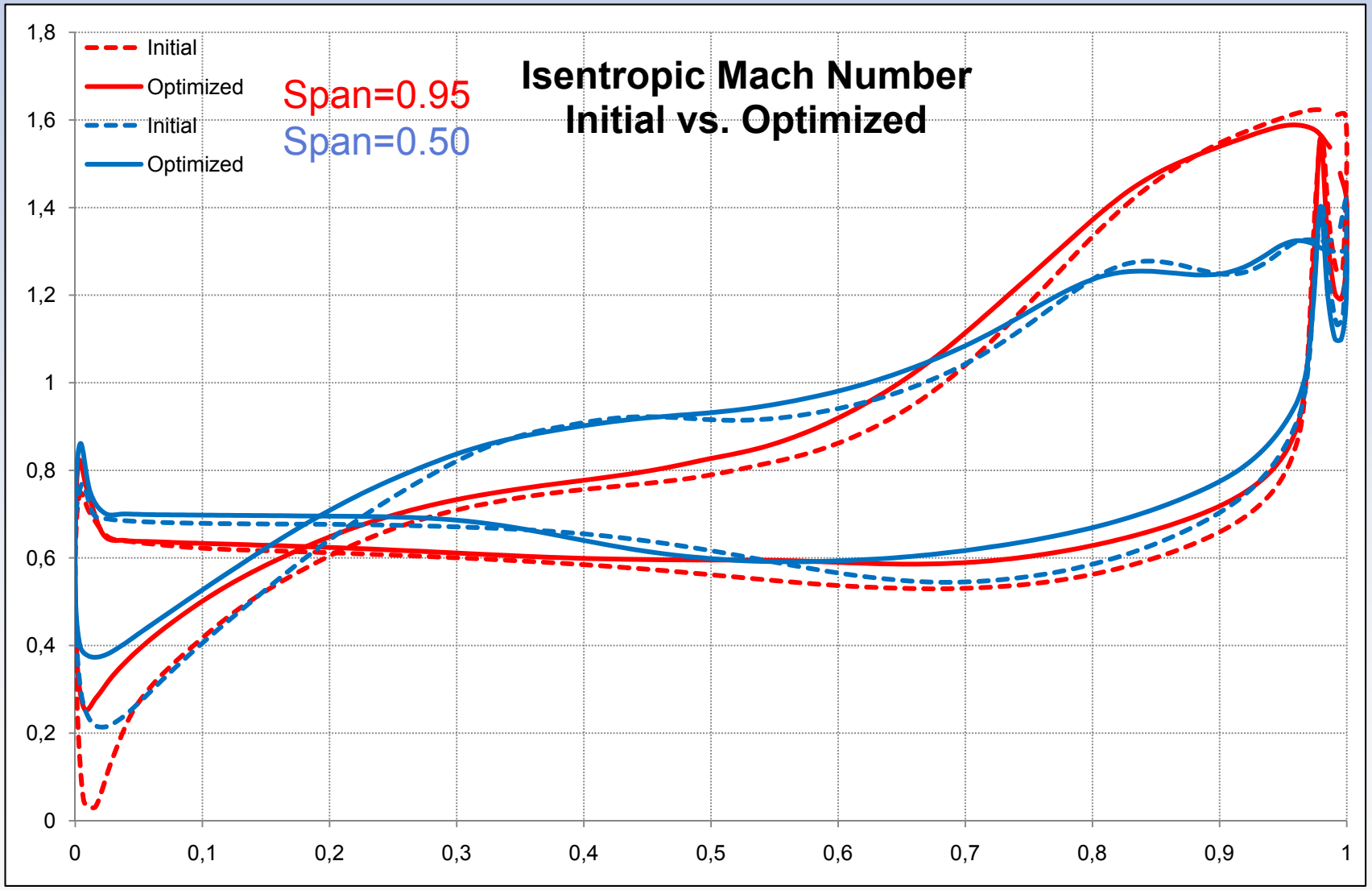
Initial vs. Optimized Design

Input	Initial Design	Optimized
α_{Hub}	0	-0.34
α_{Shroud}	0	-0.18
Ω [rev/s]	-335	-365
$\alpha_{Guide\ Vane}$	0	-9.68



Output	Initial Design	Optimized
T_t Ratio	1.116	1.151
p_t Ratio	1.674	1.848
η [%]	71.65	81.54
Power [MW]	1.208	1.329

Initial vs. Optimized Design



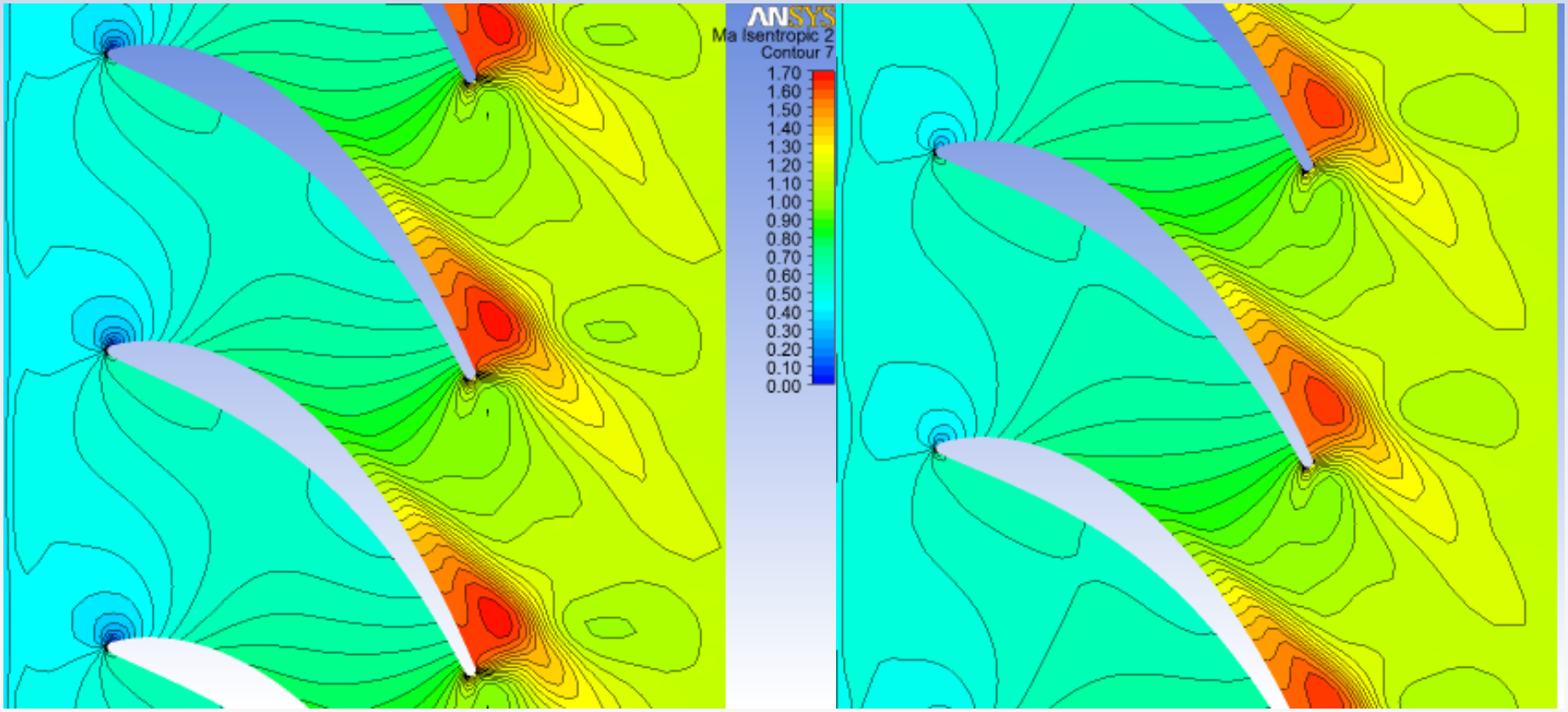
Initial vs. Optimized Design

Isotropic Mach Number

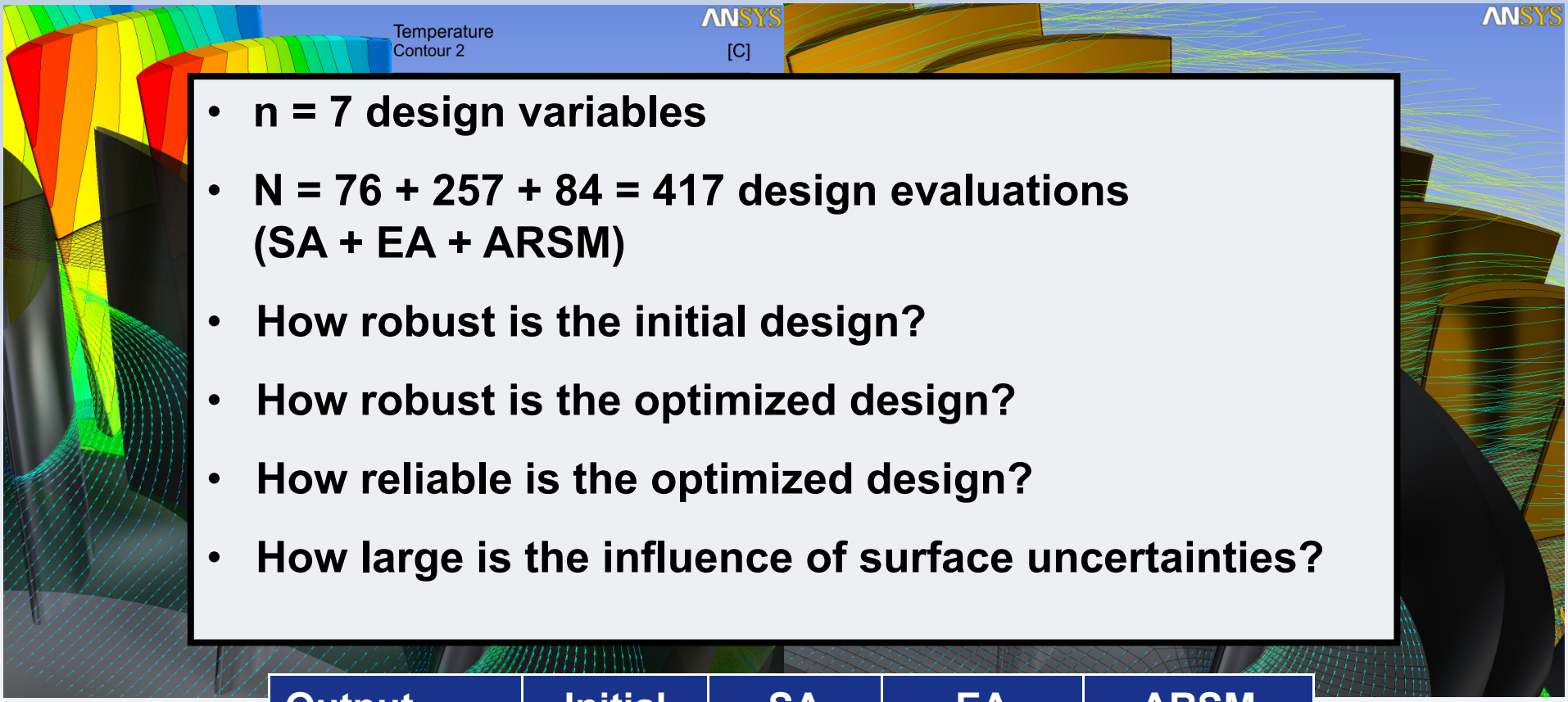
Initial

Rotor

Optimized

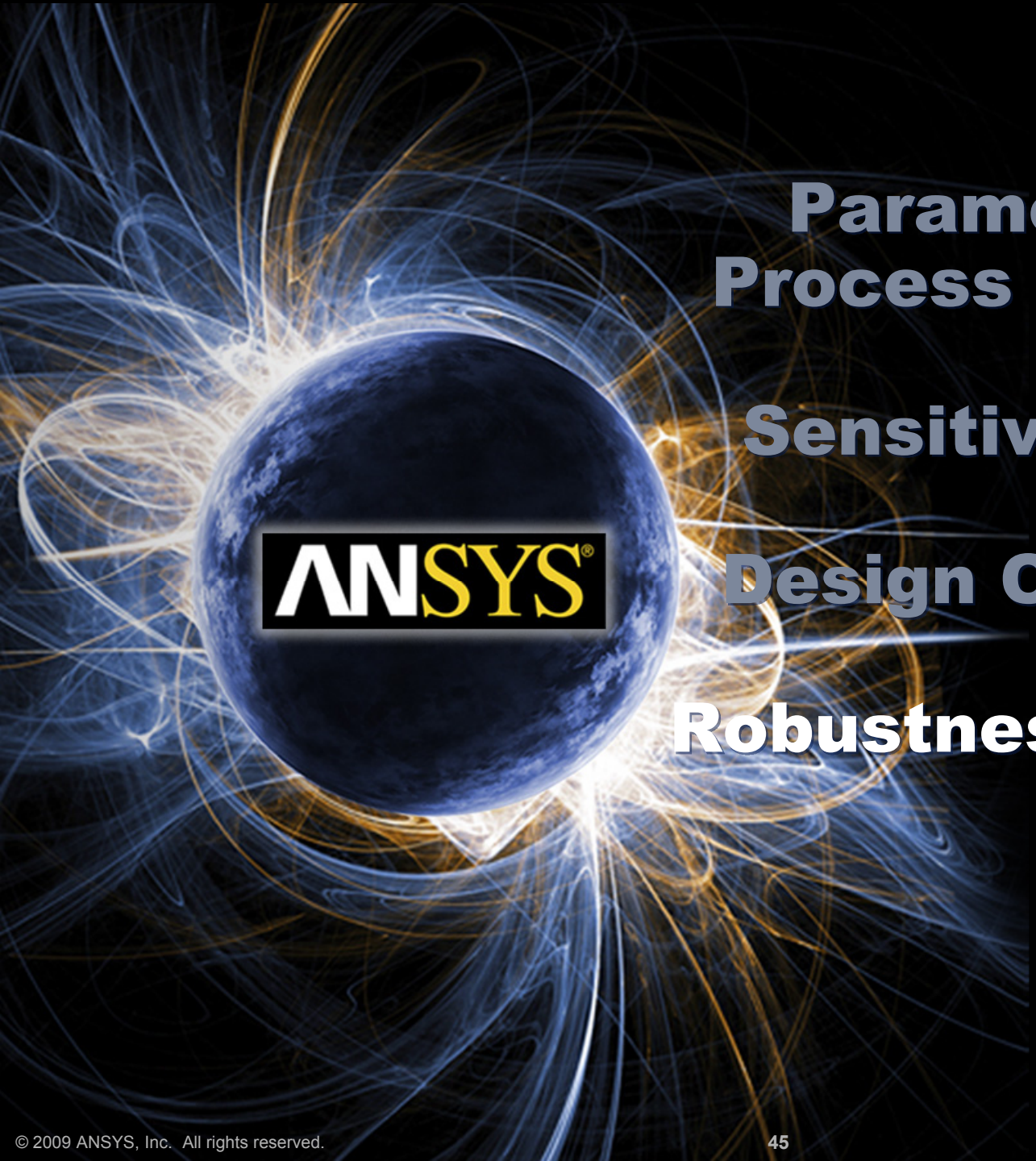


Initial vs. Optimized Design



- n = 7 design variables
- N = 76 + 257 + 84 = 417 design evaluations (SA + EA + ARSM)
- How robust is the initial design?
- How robust is the optimized design?
- How reliable is the optimized design?
- How large is the influence of surface uncertainties?

Output	Initial	SA	EA	ARSM
Objective	1.0766	0.90034	0.86841	0.86259
η [%]	71.65	80.60	81.26	81.54
Power [MW]	1.208	1.304	1.343	1.329



Parameterization
Process & Geometry

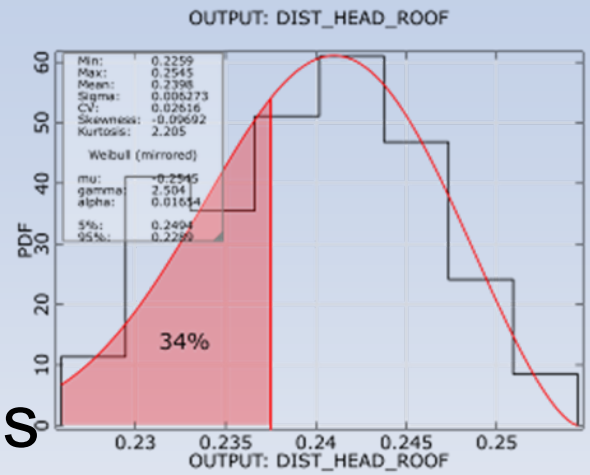
Sensitivity Analysis

Design Optimization

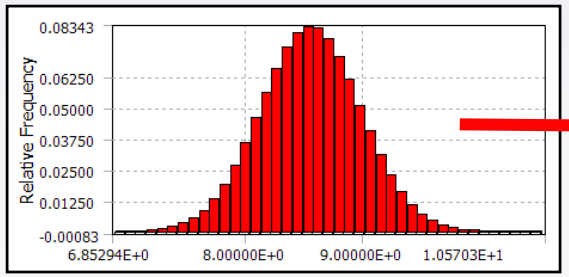
Robustness Evaluation

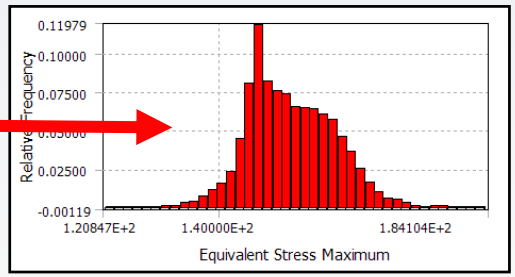
Robustness Evaluation

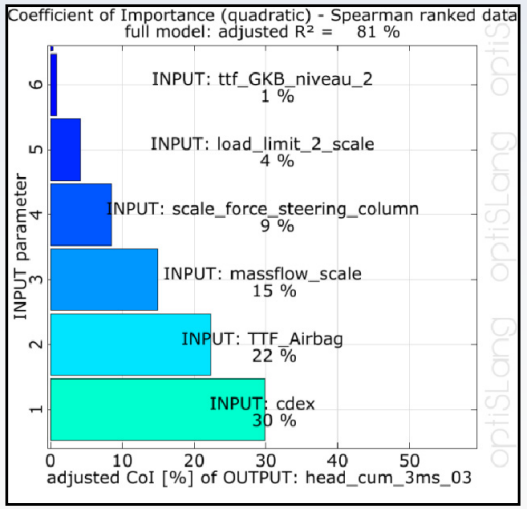
1. Scanning the Robust Space with optimized LHS, variation and correlation are investigated



2. Identification of important variables
   - Check Variation of Robust Space
   - Check Histogram, limits, probabilities
   - Check Coefficient of Importance

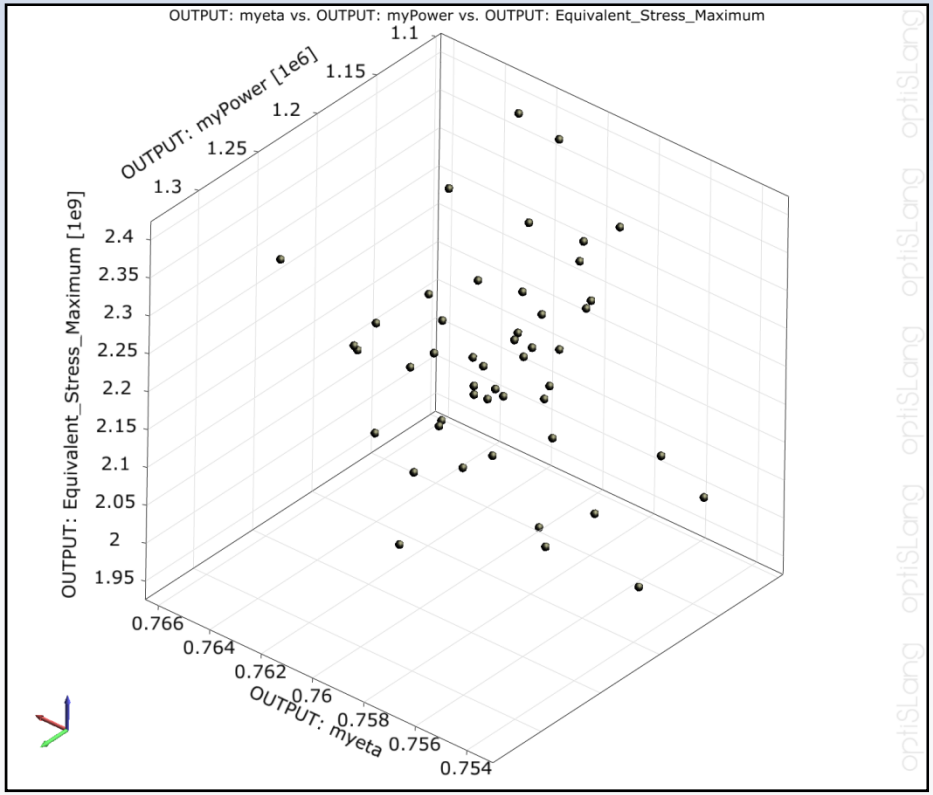


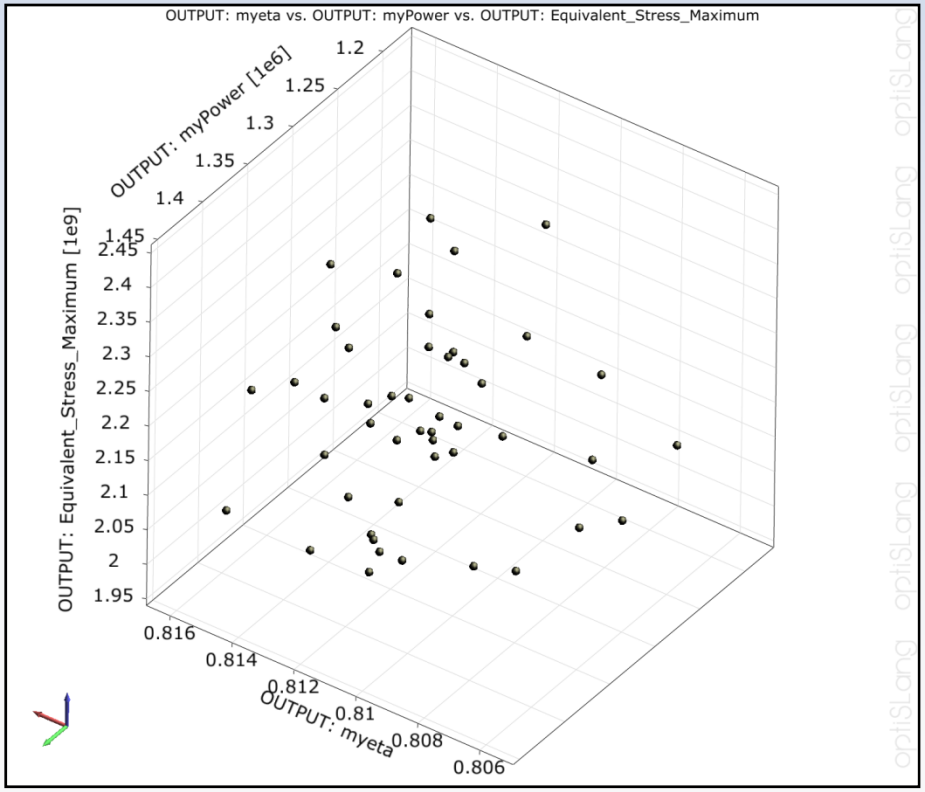




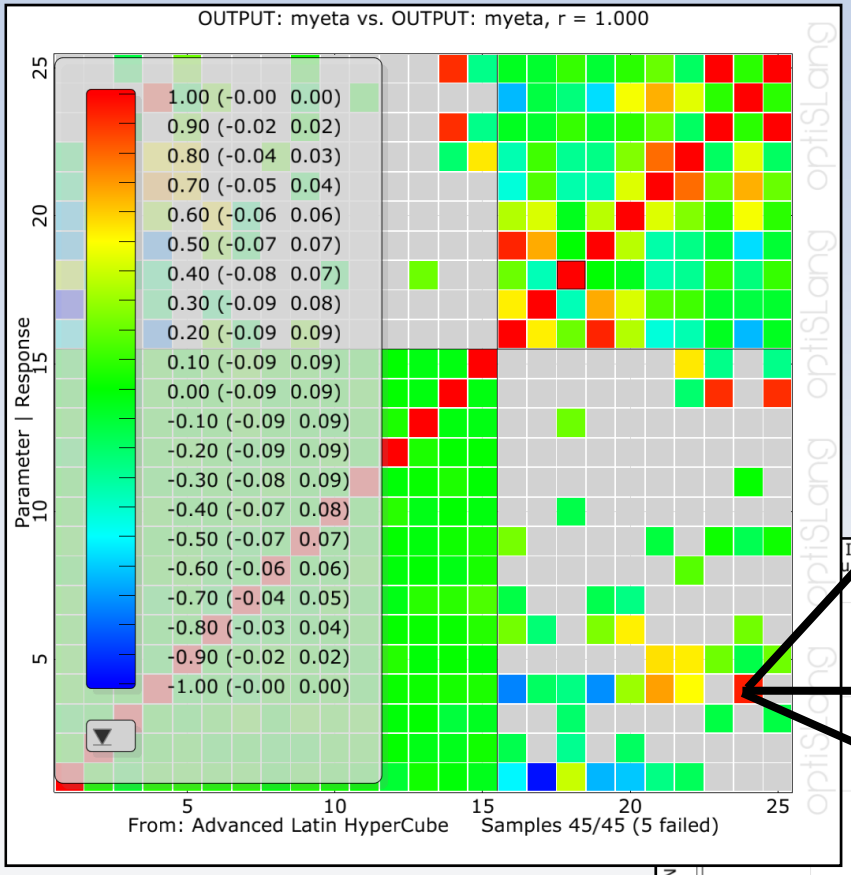
Variation of Robust Space

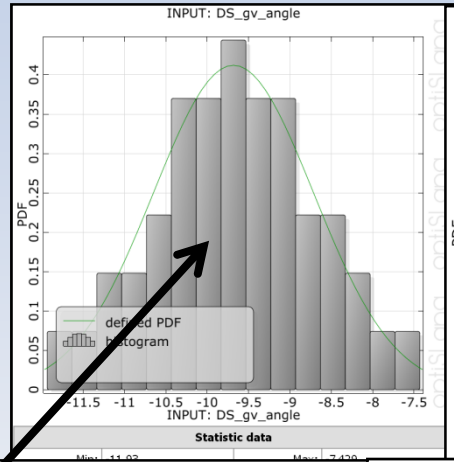
- n = 15 random parameters
- N = 50 design evaluations
- Initial vs. optimized design

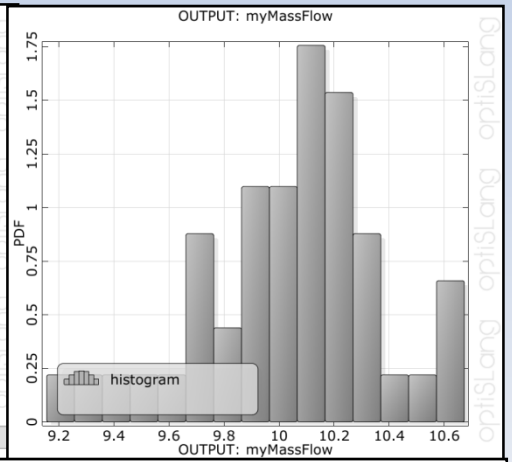


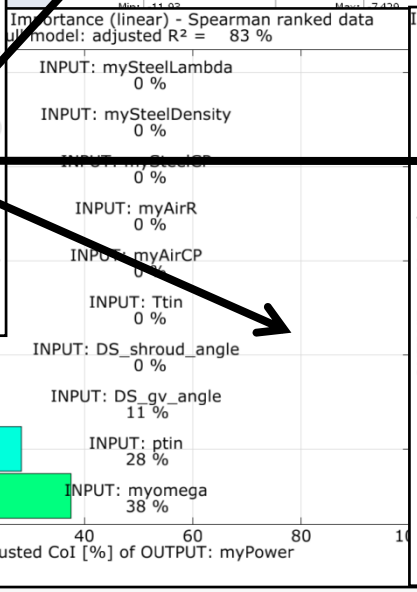


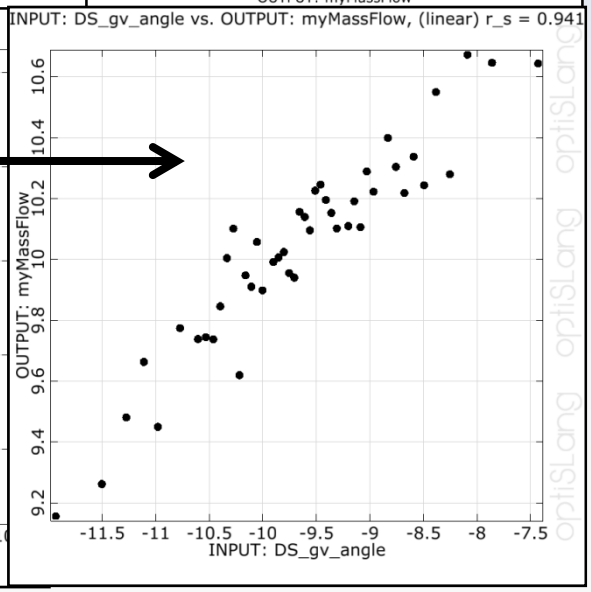
Correlation, CoI, Histogram







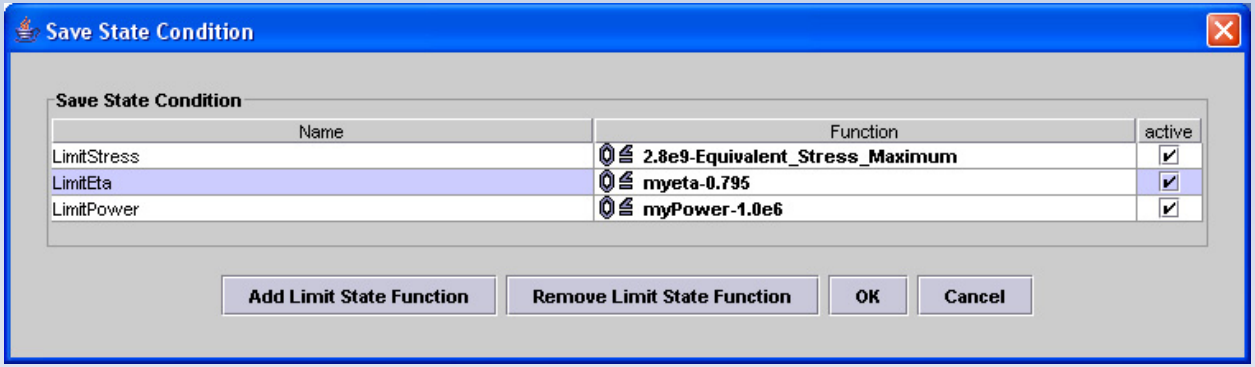




Guide vane angle vs. mass flow rate

Define Objective Limits

- Limit state conditions



- Random parameters

AxialTurbine_04ROBUSTorg.pro

Opti | Robust | Output | Strings | Constraints | Objectives

#	Name	Distribution	Mean	CoV	Std...	L...	...	...	...	
1	myomega	Normal	-2094.39	-0.02	41....	–	–	...	☑	
2	DS_hub_angle	Normal	0.0	Infinity	0.0...	–	–	...	☑	
3	DS_shroud_angle	Normal	0.0	Infinity	0.0...	–	–	...	☑	
4	DS_gv_angle	Normal	0.0	Infinity	0.1...	–	–	...	☑	
5	Ttin	Normal	1000.0	0.02	20.0	–	–	...	☑	
6	ptin	Normal	300000.0	0.03	900...	–	–	...	☑	
7	pout	Normal	87000.0	0.02	174...	–	–	...	☑	
8	myAirCP	Normal	1004.4	0.03	30....	–	–	...	☑	
9	myAirR	Normal	287.102	0.03	8.6...	–	–	...	☑	
10	mySteelCP	Normal	434.0	0.05	21....	–	–	...	☑	
11	mySteelDensity	Normal	7850.0	0.01	78.5	–	–	...	☑	
12	mySteelLambda	Normal	60.5	0.04	2.42	–	–	...	☑	
13	DS_FBlendRotor	Normal	1.0	0.02	0.02	–	–	...	☑	
14	Youngs_Modulus	Normal	2.0E11	0.03	6.0E9	–	–	...	☑	
15	Poissons_Ratio	Normal	0.3	0.1	0.03	–	–	...	☑	

Cancel OK

AxialTurbine_04ROBUST.pro

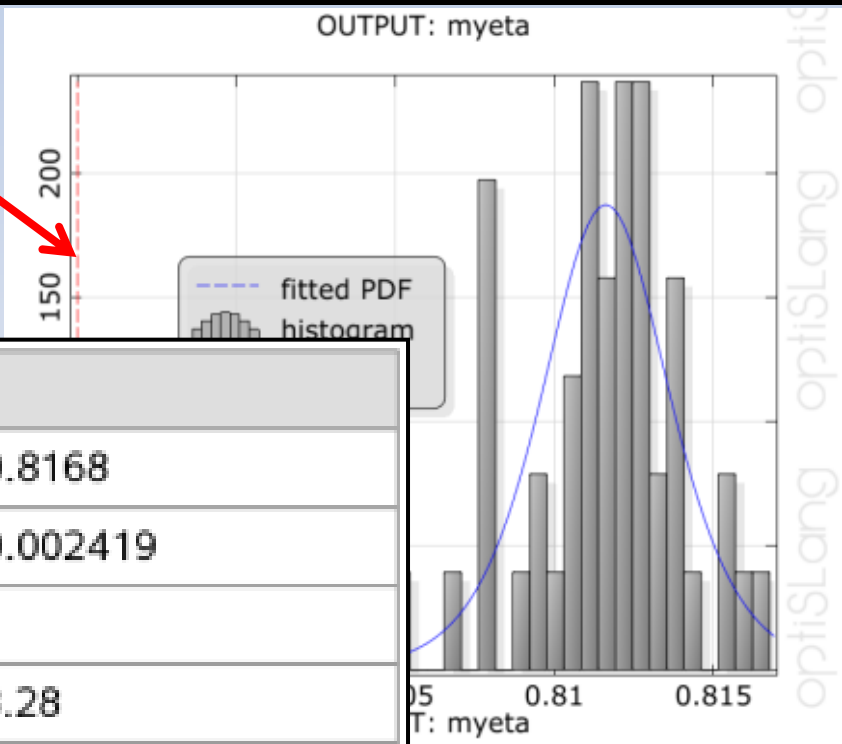
Opti | Robust | Output | Strings | Constraints | Objectives

#	Name	Distribution	Mean	CoV	Stddev	Lo...	U...	...	...	
1	myomega	Normal	-2244.21541...	-0.02	44.88...	–	–	...	☑	
2	DS_hub_angle	Normal	-0.33698805...	-0.02	0.006...	–	–	...	☑	
3	DS_shroud_angle	Normal	-0.17857451...	-0.02	0.003...	–	–	...	☑	
4	DS_gv_angle	Normal	-9.68181702...	-0.02	0.193...	–	–	...	☑	
5	Ttin	Normal	1002.859199...	0.02	20.05...	–	–	...	☑	
6	ptin	Normal	305000.0	0.03	9150.0	–	–	...	☑	
7	pout	Normal	87000.0	0.02	1740.0	–	–	...	☑	
8	myAirCP	Normal	1004.4	0.03	30.13...	–	–	...	☑	
9	myAirR	Normal	287.102	0.03	8.613...	–	–	...	☑	
10	mySteelCP	Normal	434.0	0.05	21.70...	–	–	...	☑	
11	mySteelDensity	Normal	7850.0	0.01	78.5	–	–	...	☑	
12	mySteelLambda	Normal	60.5	0.04	2.42	–	–	...	☑	
13	DS_FBlendRotor	Normal	0.9	0.02	0.018...	–	–	...	☑	
14	Youngs_Modulus	Normal	2.0E11	0.03	6.0E9	–	–	...	☑	
15	Poissons_Ratio	Normal	0.3	0.1	0.03	–	–	...	☑	

Cancel OK

Evaluation of Histogram

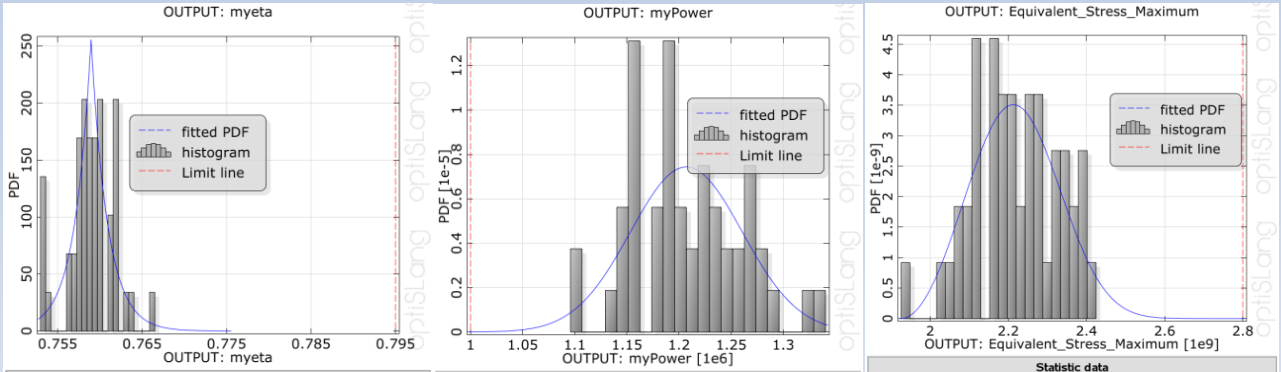
Limit for Variable



Statistic data			
Min:	0.8049	Max:	0.8168
Mean:	0.8116	Sigma:	0.002419
CV:	0.00298		
Skewness:	-0.4368	Kurtosis:	3.28
Fitted PDF: Logistic			
Mean:	0.8116	Sigma:	0.002419
Limit x = 0.795			
P_rel =	0	P_fit =	3.87374e-06

tic data			
		Max:	0.8168
		Sigma:	0.002419
		Kurtosis:	3.28
DF: Logistic			
		Sigma:	0.002419
Limit x = 0.795			
P_rel =	0	P_fit =	3.87374e-06

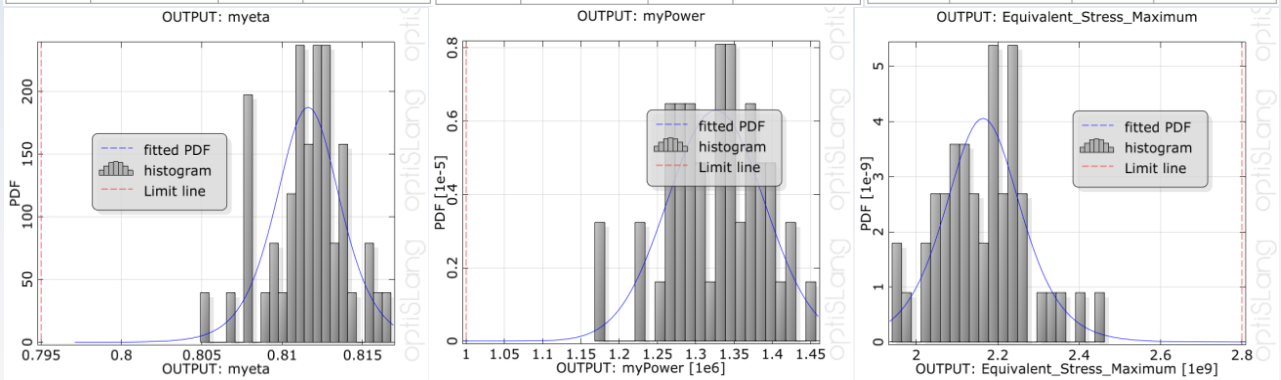
Evaluation of Histogram



Statistic data			
Min:	0.753	Max:	0.7665
Mean:	0.759	Sigma:	0.002767
CV:	0.003645		
Skewness:	-0.1759	Kurtosis:	3.679
Fitted PDF: Laplace			
Mean:	0.759	Sigma:	0.002767
Limit x = 0.795			
P_rel =	1	P_fit =	1

Statistic data			
Min:	1.096e+06	Max:	1.341e+06
Mean:	1.207e+06	Sigma:	5.364e+04
CV:	0.04444		
Skewness:	0.2879	Kurtosis:	2.766
Fitted PDF: Normal			
Mean:	1.207e+06	Sigma:	5.364e+04
Limit x = 1e+06			
P_rel =	0	P_fit =	5.67009e-05

Statistic data			
Min:	1.926e+09	Max:	2.424e+09
Mean:	2.221e+09	Sigma:	1.086e+08
CV:	0.04892		
Skewness:	-0.2418	Kurtosis:	2.725
Fitted PDF: Extreme Typ III (Max) Weibull			
Mean:	2.221e+09	Sigma:	1.086e+08
Lower cut:	1.926e+09		
Limit x = 2.8e+09			
P_rel =	1	P_fit =	1



Statistic data			
Min:	0.8049	Max:	0.8168
Mean:	0.8116	Sigma:	0.002419
CV:	0.00298		
Skewness:	-0.4368	Kurtosis:	3.28
Fitted PDF: Logistic			
Mean:	0.8116	Sigma:	0.002419
Limit x = 0.795			
P_rel =	0	P_fit =	3.87374e-06

Statistic data			
Min:	1.168e+06	Max:	1.457e+06
Mean:	1.326e+06	Sigma:	6.335e+04
CV:	0.04776		
Skewness:	-0.3664	Kurtosis:	2.825
Fitted PDF: Normal			
Mean:	1.326e+06	Sigma:	6.335e+04
Limit x = 1e+06			
P_rel =	0	P_fit =	1.3062e-07

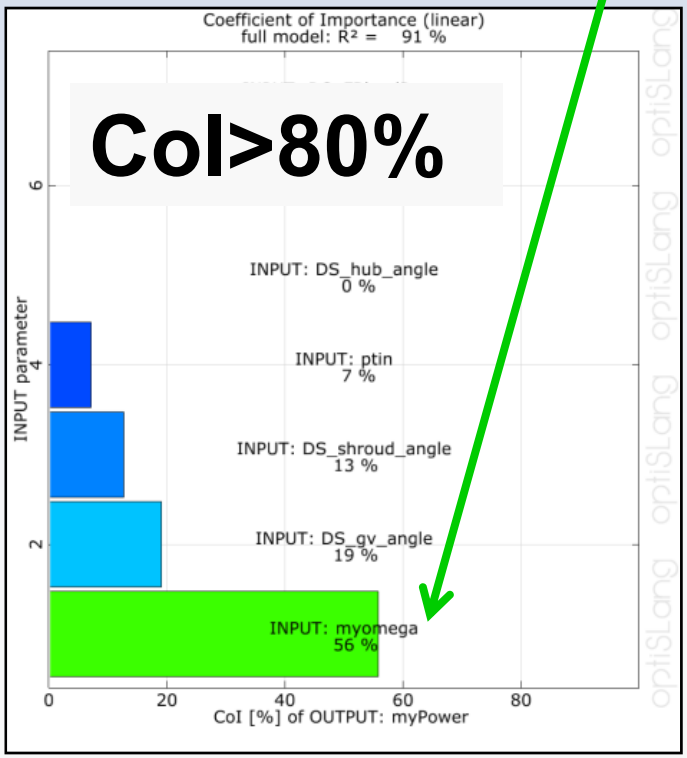
Statistic data			
Min:	1.94e+09	Max:	2.462e+09
Mean:	2.164e+09	Sigma:	1.118e+08
CV:	0.05165		
Skewness:	0.3229	Kurtosis:	3.244
Fitted PDF: Logistic			
Mean:	2.164e+09	Sigma:	1.118e+08
Limit x = 2.8e+09			
P_rel =	1	P_fit =	0.999967

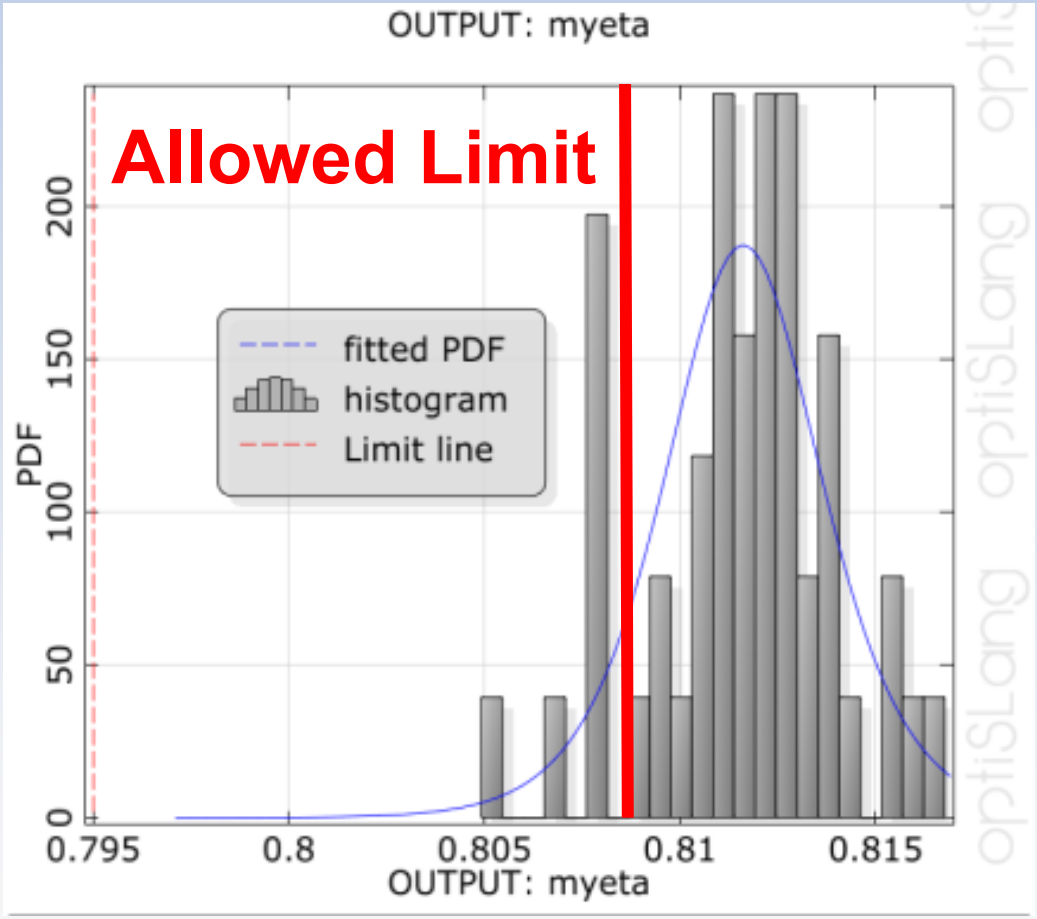
Non-robust initial design

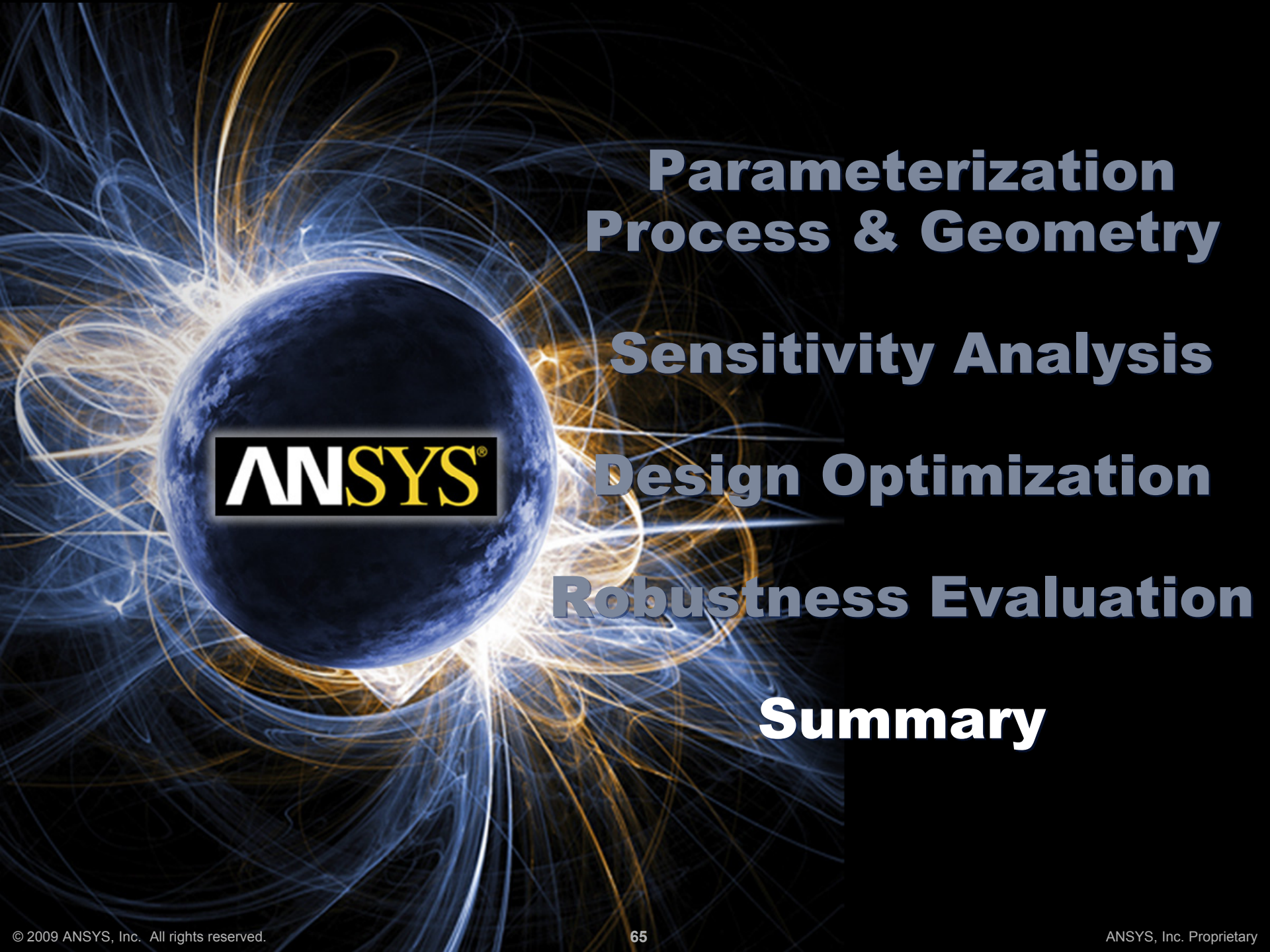
Robust optimized design up to a sigma level of 4.5

Conclusion Robustness Evaluation

- Is there a problem?
- Can we explain, CoI?
- Who is responsible?







Parameterization
Process & Geometry

Sensitivity Analysis

Design Optimization

Robustness Evaluation

Summary

Summary

- **Workbench supports full Workflow**
  - Geometry, Meshing, Simulation, Post-Processing
- **Multi Physics support**
- **Parametric Workflow management**
- **Automatic and embedded solution procedure**
- **Sensitivity Analysis**
- **Design Optimization**
- **Robustness Evaluation**