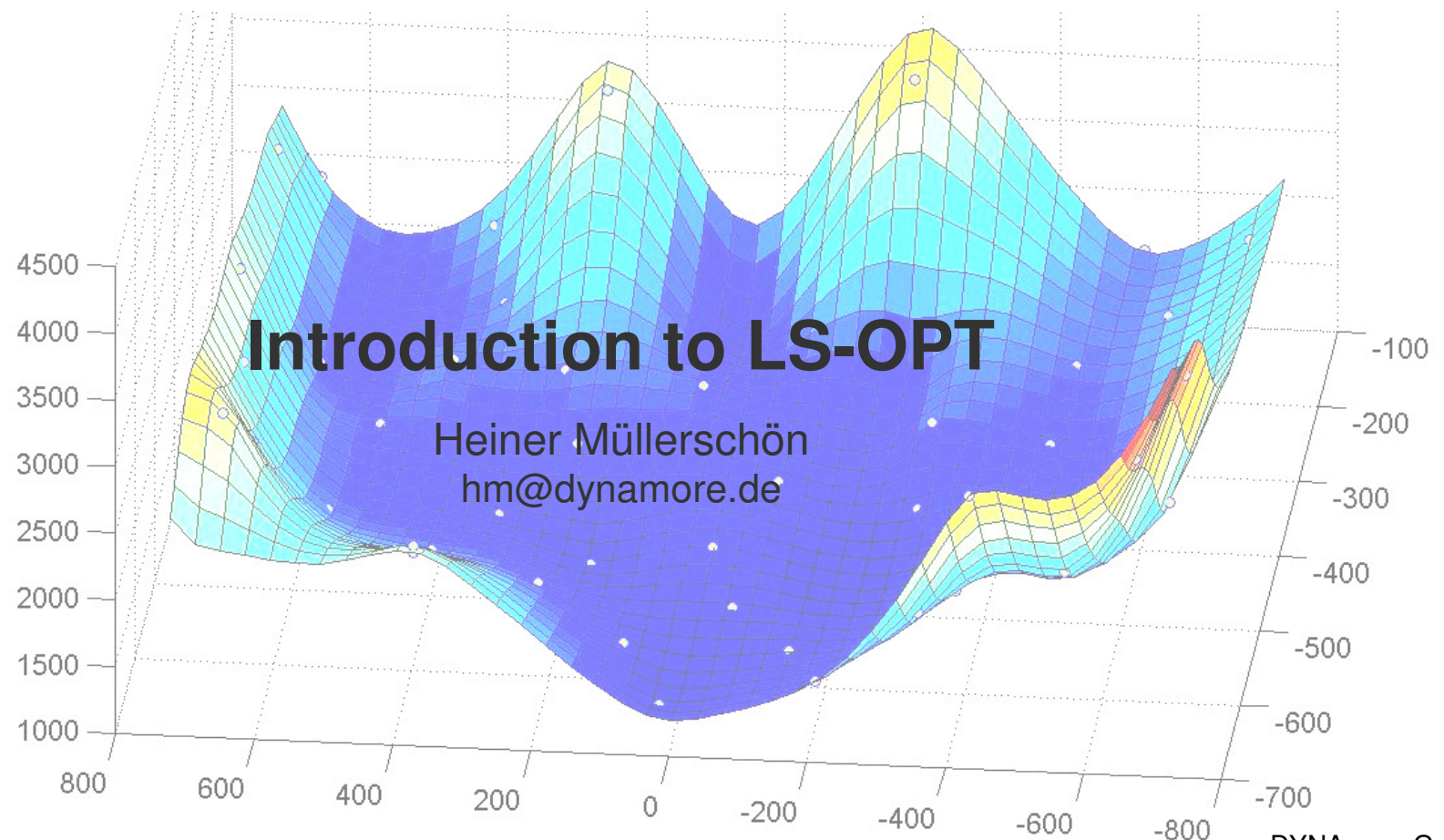




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Introduction to LS-OPT

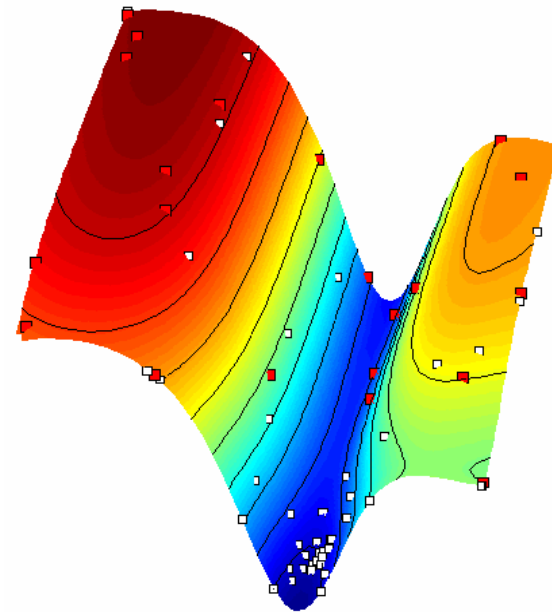
Heiner Müllerschön
hm@dynamore.de

DYNAmore GmbH
Industriestraße 2
70565 Stuttgart
<http://www.dynamore.de>



┌ Overview

- n Introduction/Features
- n Methodologies – Optimization
- n Methodologies - Robustness
- n Examples - Optimization
- n Examples - Robustness
- n Version 3.3 / Outlook



Introduction / Features

- Introduction/Features
- Methods – Optimization
- Methods - Robustness
- Examples - Optimization
- Examples - Robustness
- Version 3.3 / Outlook

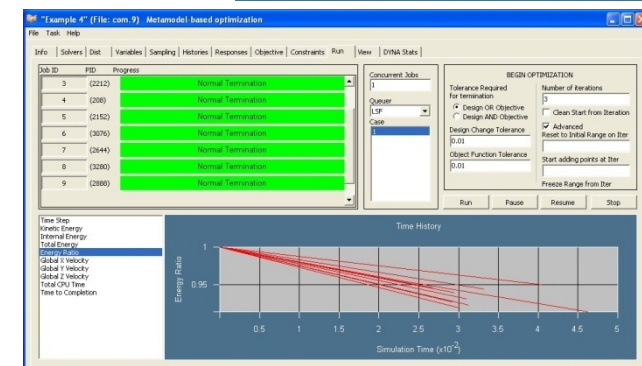
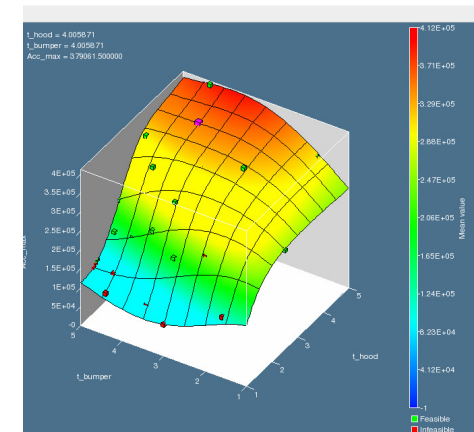


About LS-OPT

- LS-OPT is a **product of LSTC** (Livermore Software Technology Corporation)
- LS-OPT can be linked to any **simulation code** – stand alone optimization software

Methodologies/Features:

- Successive Response Surface Method (SRSM)
- Genetic Algorithm (MOGA->NSGA-II)
- Multidisciplinary optimization (MDO)
- Multi-Objective optimization (Pareto)
- numerical/analytical based sensitivities
- Analysis of Variance (ANOVA)
- Stochastic/Probabilistic Analysis
- Monte Carlo Analysis using Metamodels
-



About LS-OPT

Mixed Discrete-Continuous Optimization

- Specify sets of discrete variables (e.g. sheet thicknesses)

Robust Parameter Design (RDO)

- Improve/Maximizing the robustness of the optimum

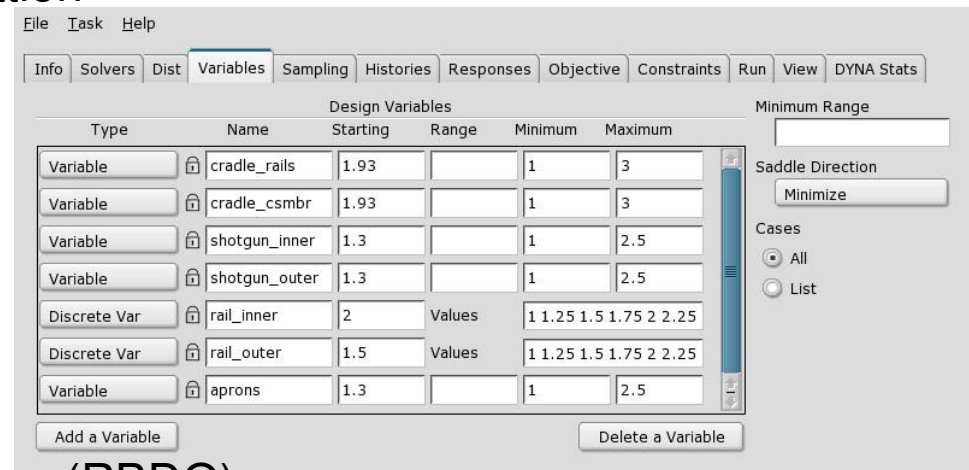
Reliability Based Design Optimization (RBDO)

- Improve failure probability of optimum

Visualization of Stochastic Results

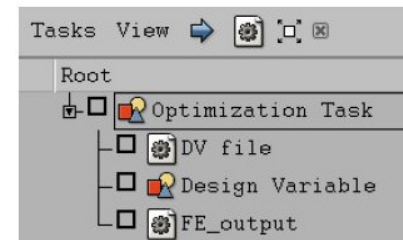
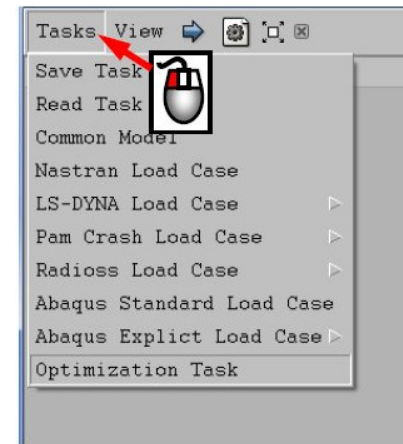
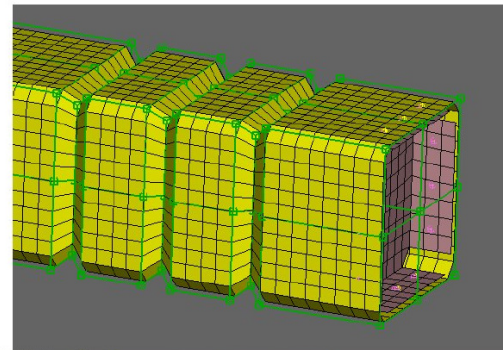
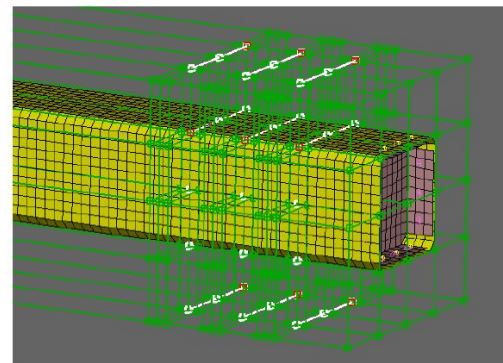
- Confidence Intervals, reliability quantities

- Fringe of statistic results on the FE-Model



About LS-OPT

- Job Distribution - Interface to Queuing Systems
 - PBS, LSF, LoadLeveler, SLURM, AQS, etc.
 - Retry of failed queuing (abnormal termination)
- LS-OPT might be used as a “Process Manager”
- Shape Optimization
 - Interface to **ANSA**, *HyperMorph*, *DEP-Morpher*, *SFE-Concept*
 - User-defined interface to any Pre-Processor



Introduction / Features

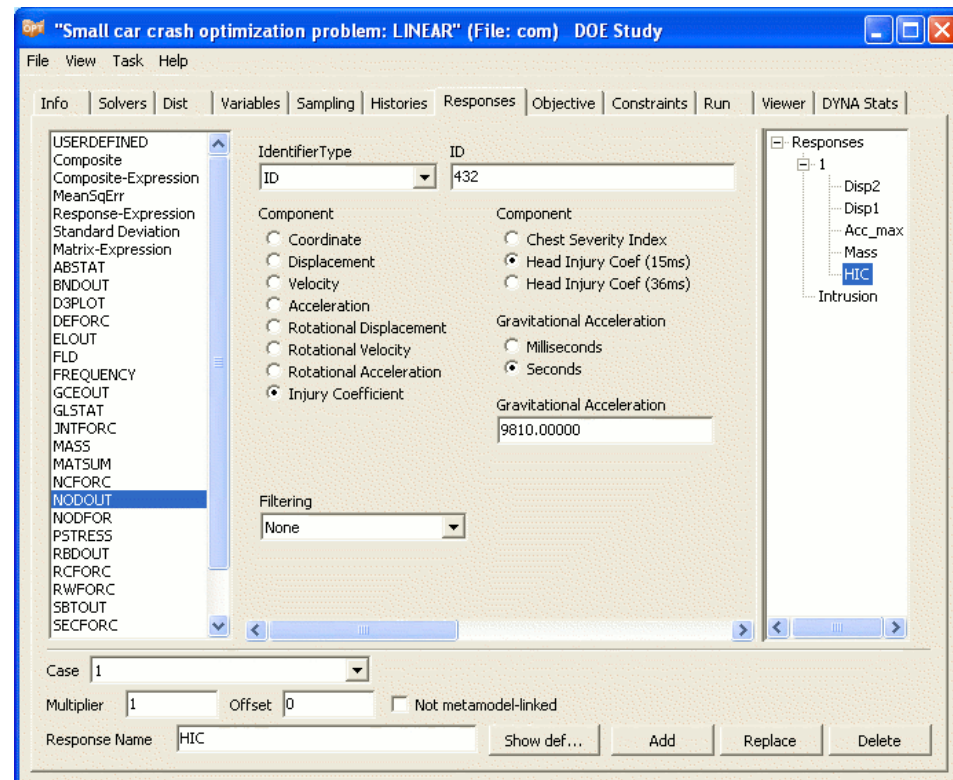
- § Introduction/Features
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- § Methods - Robustness
- § Examples - Optimization
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- § Version 3.3 / Outlook



About LS-OPT

■ LS-DYNA Integration

- *Checking of Dyna keyword files (*DATABASE_)*
- *Importation of design parameters from Dyna keyword files (*PARAMETER_)*
- *Monitoring of LS-DYNA progress*
- *Result extraction of most LS-DYNA response types*
- *D3plot compression (node and part selection)*

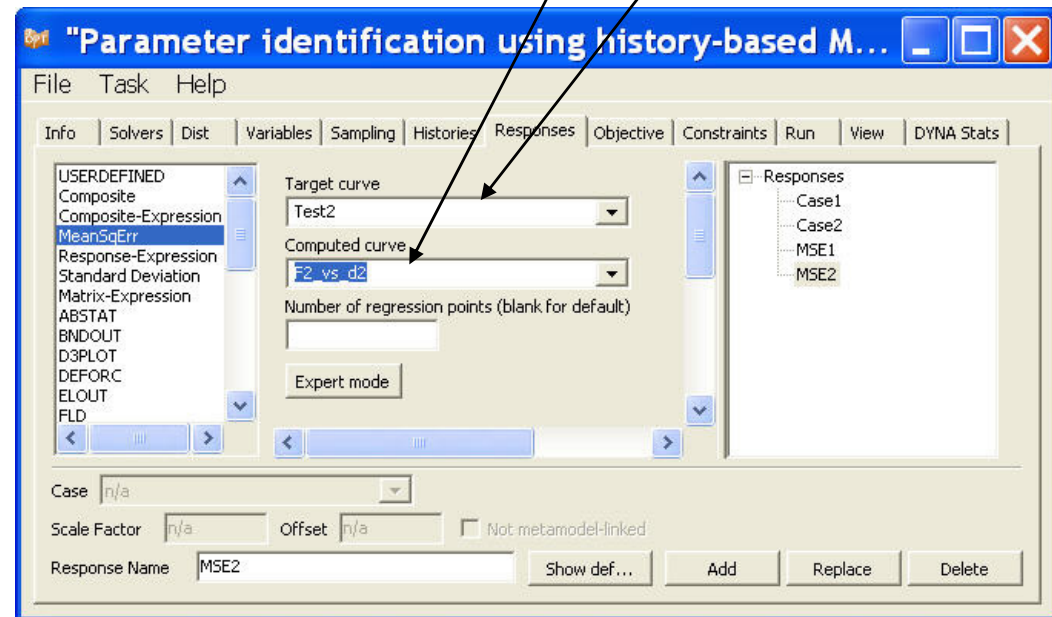


About LS-OPT

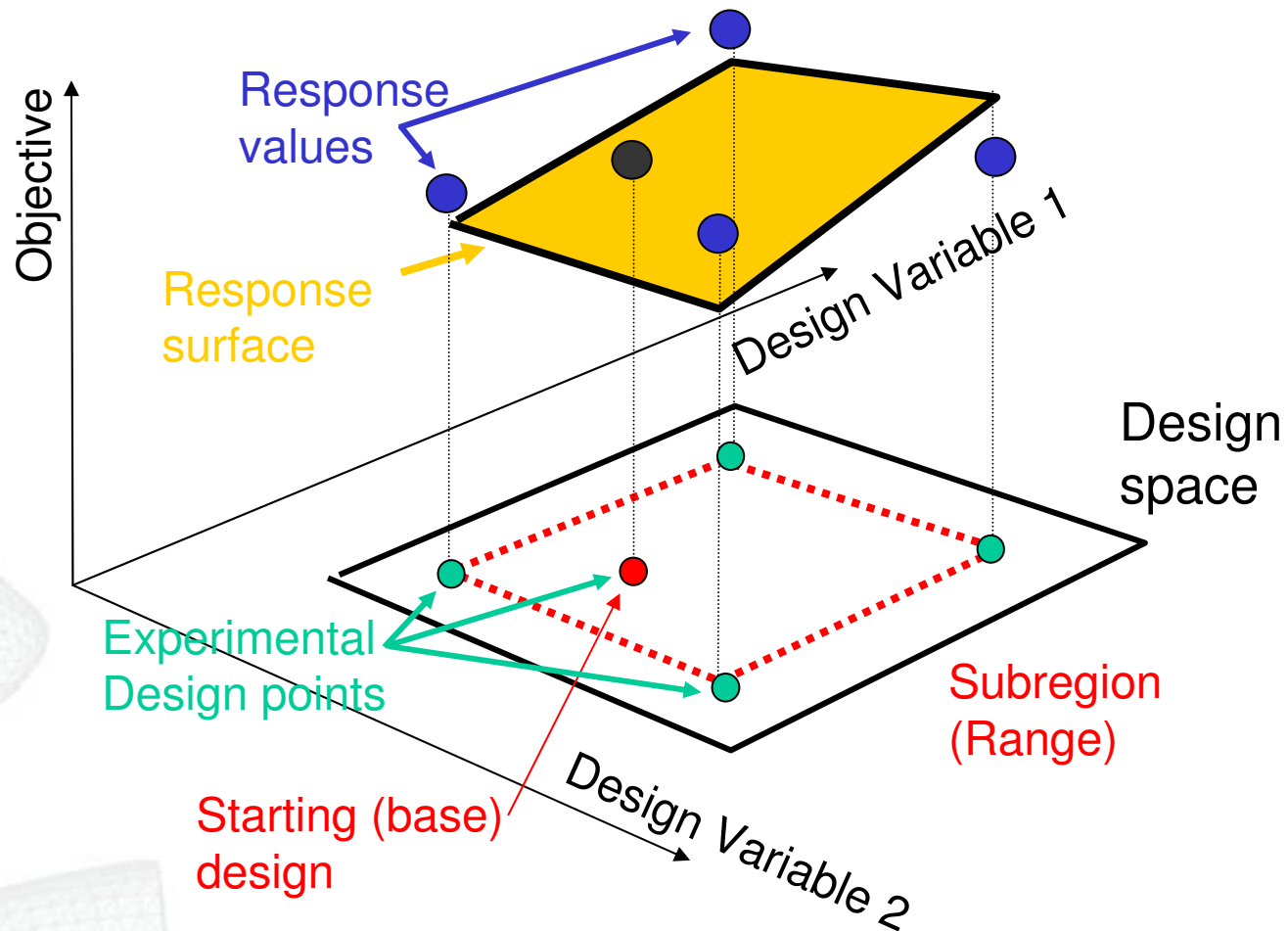
Parameter Identification Module

- Handles "continuous" test curves
- Automated use of test results to calibrate materials/systems
- Simplify input for system identification applications
- Visualization of test and simulation curve to compare
- Confidence intervals for individual parameters in parameter identification

$$\frac{1}{P} \sum_{p=1}^P W_i \left(\frac{F_i(\mathbf{x}) - G_i}{s_i} \right)^2$$



↳ Response Surface Methodology - Optimization Process

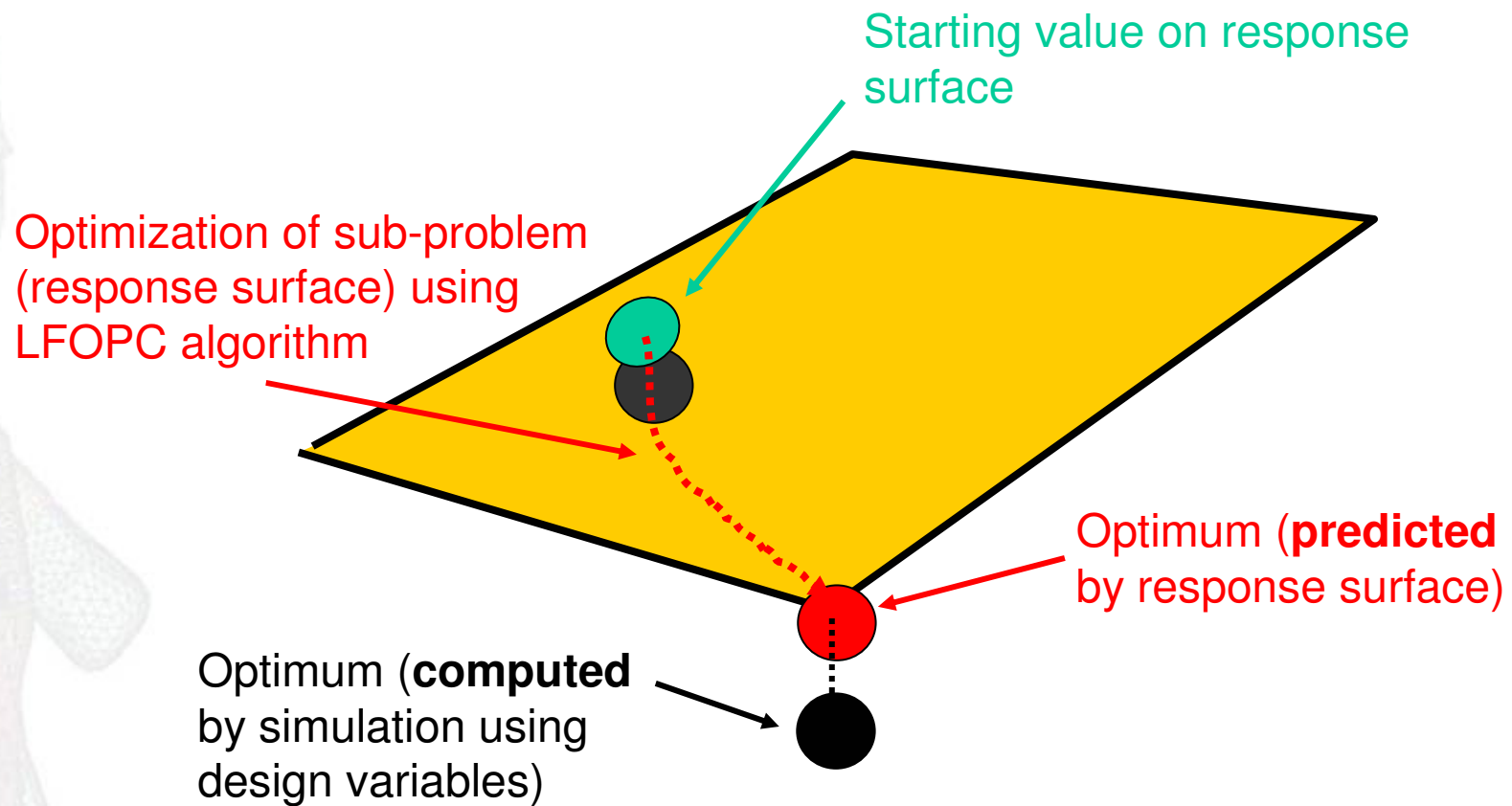


Methods - Optimization

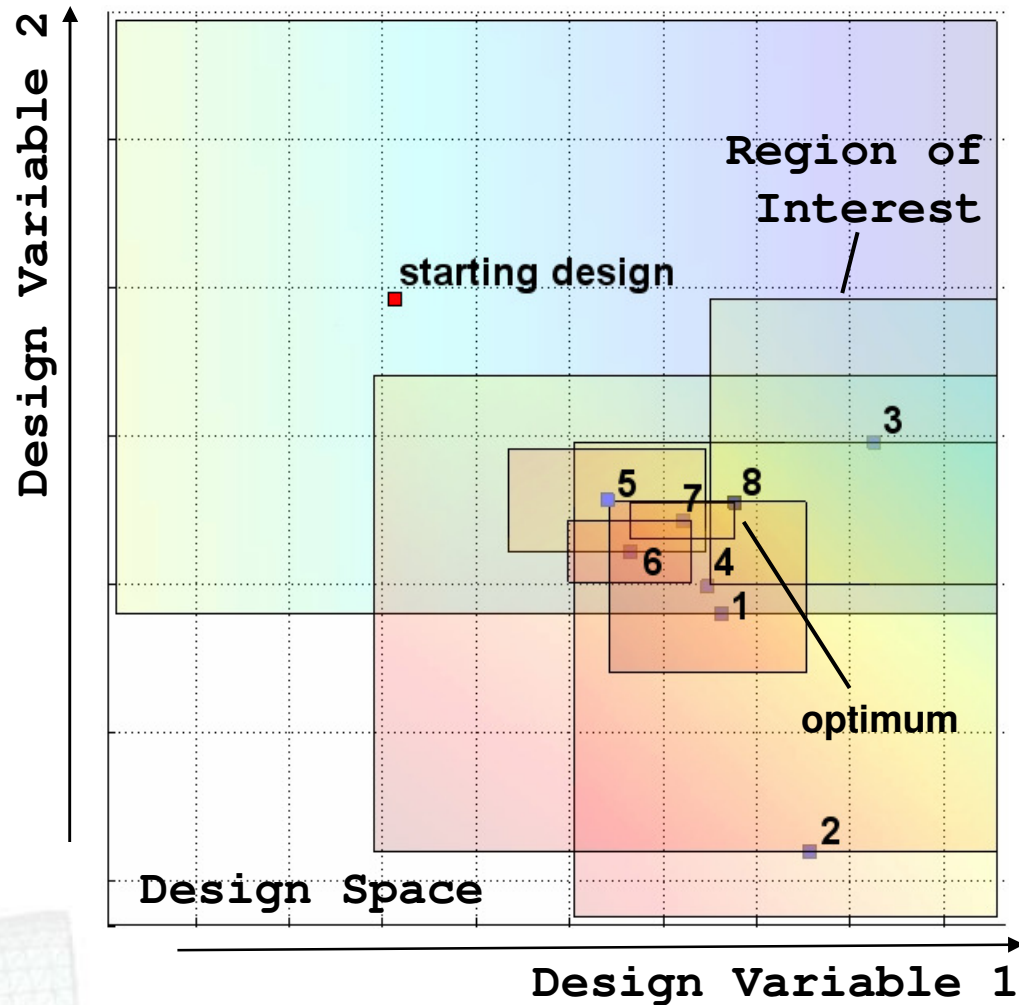
- Introduction/Features
- Methods - Optimization**
- Methods - Robustness
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Find an Optimum on the Response Surface (one iteration)



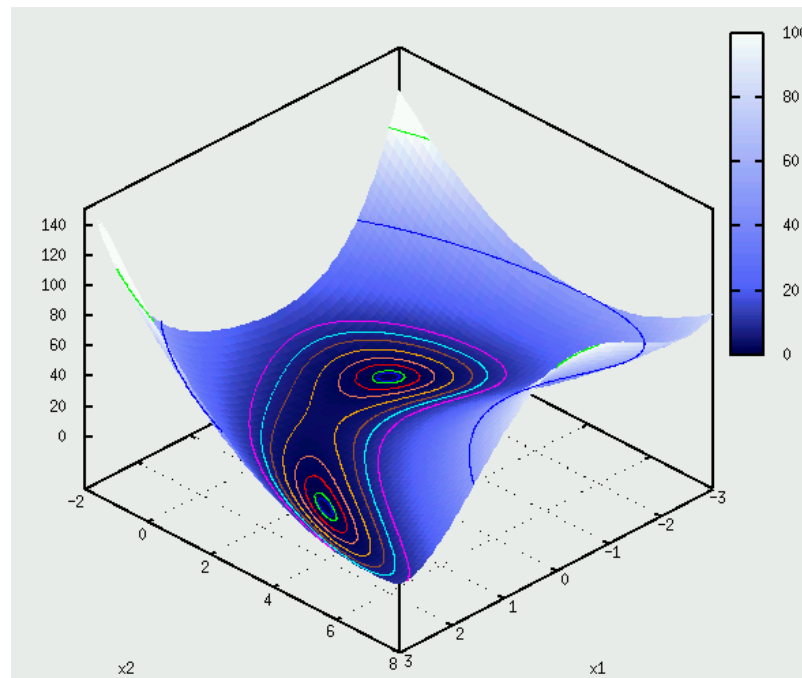
Successive Response Surface Methodology



Successive Response Surface Methodology

■ Example - 4th order polynomial

$$g(\mathbf{x}) = 4 + \frac{9}{2}x_1 - 4x_2 + x_1^2 + 2x_2^2 - 2x_1x_2 + x_1^4 - 2x_1^2x_2$$



[movie](#)

Methods - Optimization

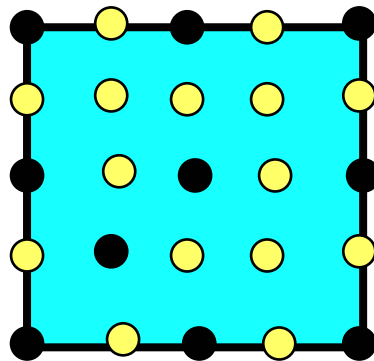
- § Introduction/Features
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- § Methods - Robustness
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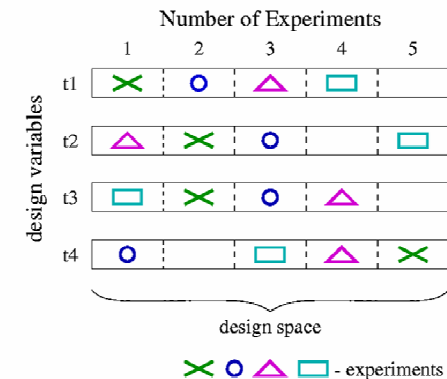
⌘ Design of Experiments (DOE) - Sampling Point Selection

- Koshal, Central Composite, Full Factorial
- **D-Optimality Criterion** - Gives maximal confidence in the model

$$\max |X^T X|$$



- Monte Carlo Sampling
- Latin Hypercube Sampling (stratified Monte Carlo)
- Space Filling Designs
- User Defined Experiments



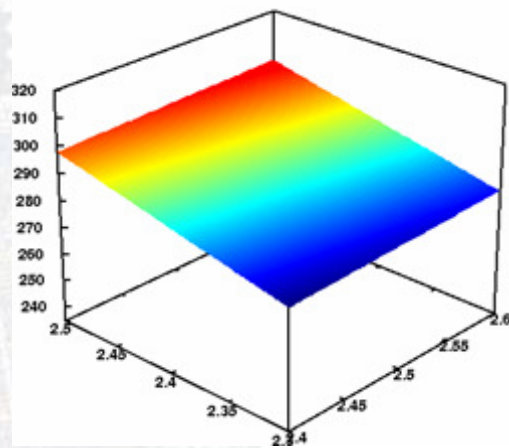
Methods - Optimization

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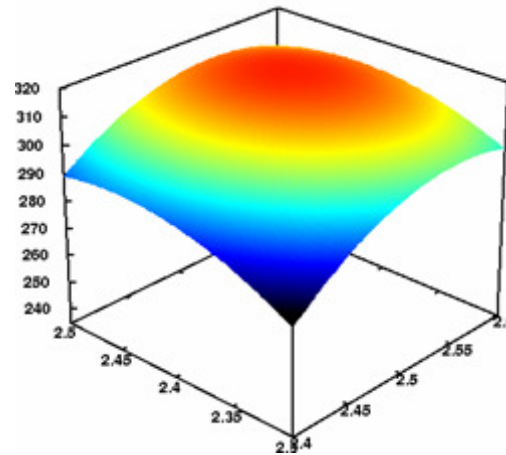


↳ Response Surfaces (Meta Models)

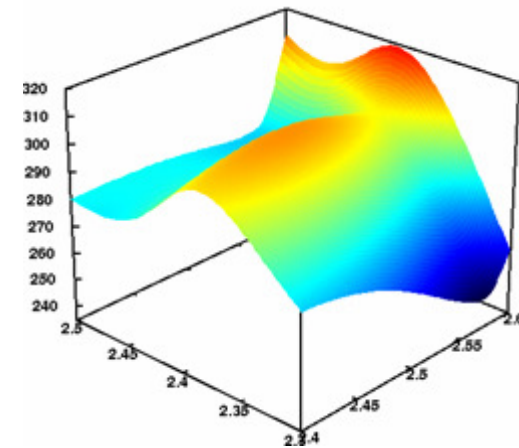
- Linear, Quadratic and Mixed polynomial based
- Neural Network and Kriging for Nonlinear Regression



linear polynomial



quadratic polynomial

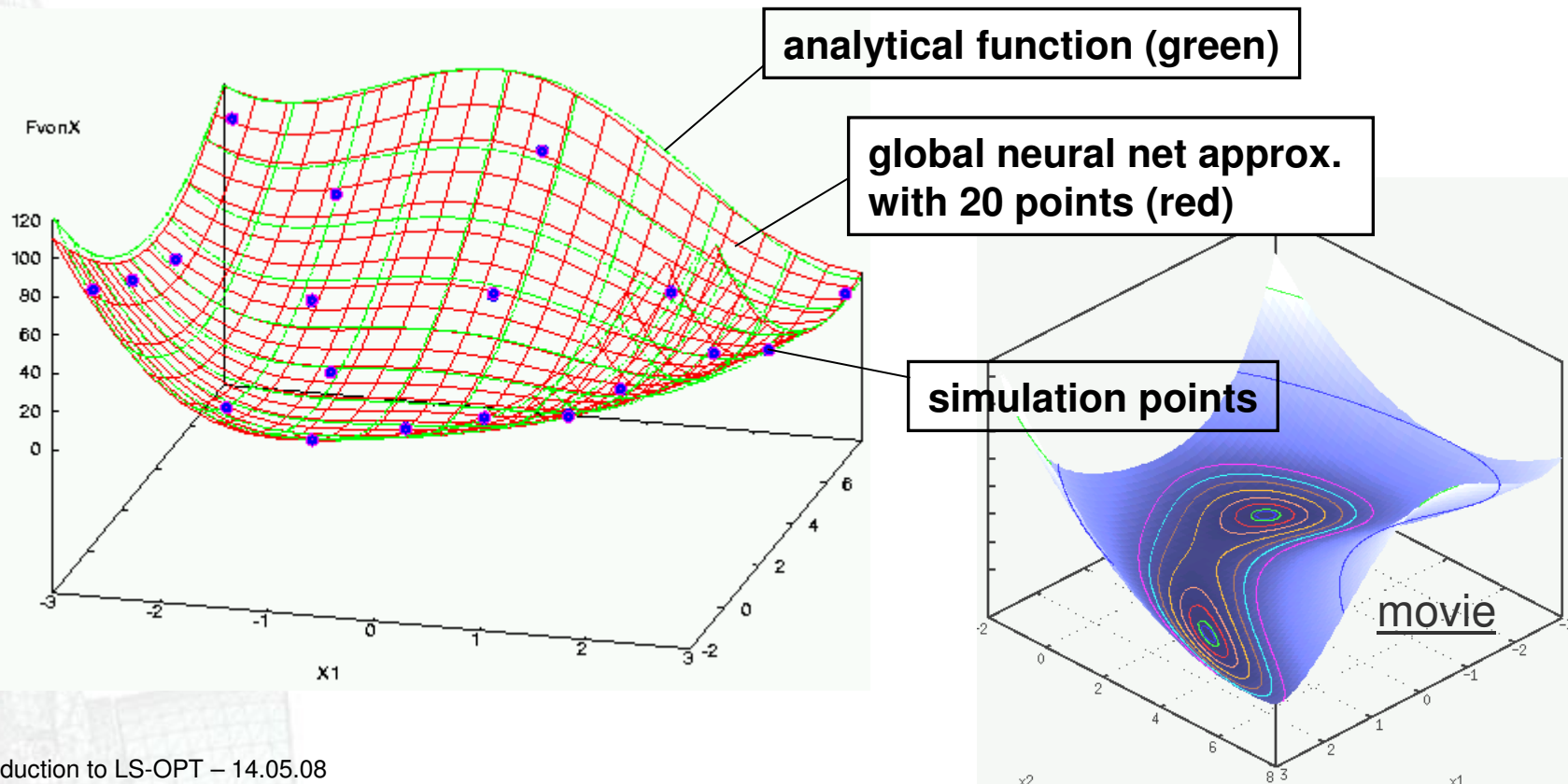


neural network

Neural Network Regression

Example - 4th order polynomial

$$g(\mathbf{x}) = 4 + \frac{9}{2}x_1 - 4x_2 + x_1^2 + 2x_2^2 - 2x_1x_2 + x_1^4 - 2x_1^2x_2$$



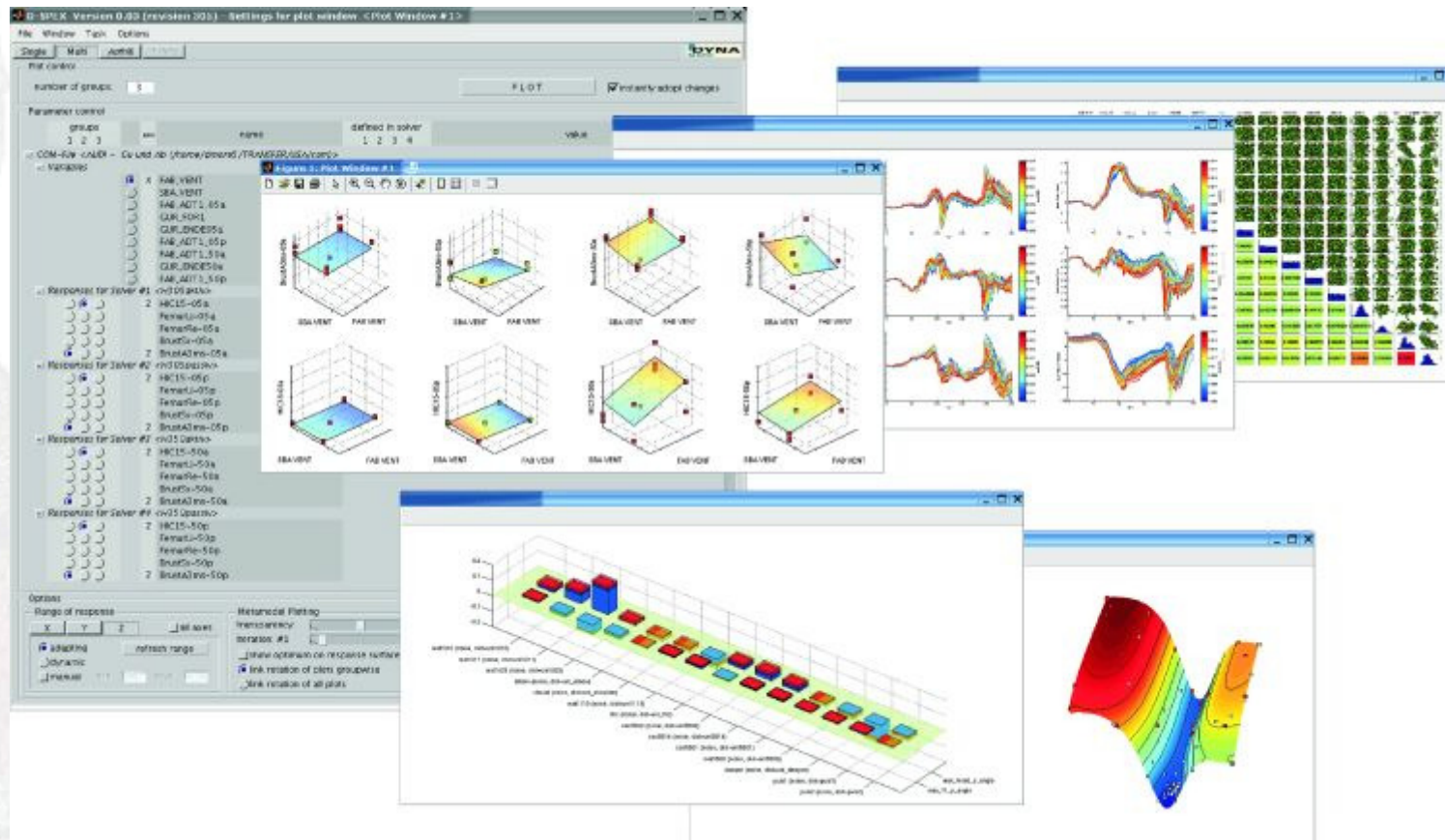
Exploring Design Space using D-SPEX

- Introduction/Features
- Methods – Optimization**
- Methods - Robustness
- Examples - Optimization
- Examples - Robustness
- Version 3.2 / Outlook

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Meta-Model Viewer - Exploration of Design Space

Compare responses, histories or even different optimization projects



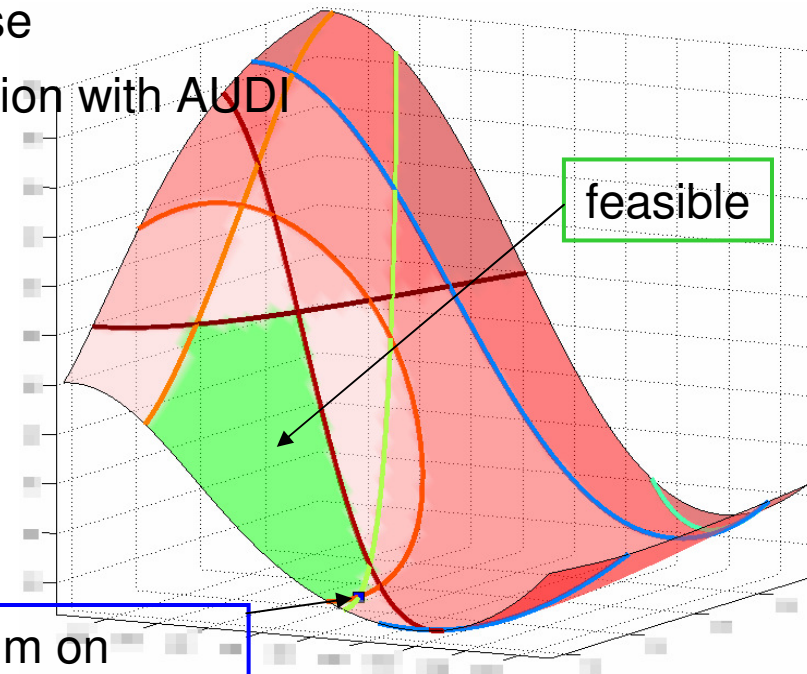
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About D-SPEX

- D-SPEX – **D**esign **S**pace **E**Xplorer
- D-SPEX is a software tool for the visualization of Meta-Models and results of optimization or stochastic analysis
- Versions Windows 32/64bit and Linux 32/64bit
- Complete interface to LS-OPT database
- Developed by DYNAmore in collaboration with AUDI (property of DYNAmore)
- Methodologies/Features:
 - *Meta-Model viewer*
 - *Curve statistics*
 - *Feasible/Infeasible design*
 - *Ant-Hill plots*
 - *Statistic evaluations*



⌘ Overview – Optimization Methodologies for highly nonlinear Applications

	Gradient Based Methods	Random Search	Genetic Algorithms	RSM / SRSM
	<ul style="list-style-type: none"> § accuracy of solution § number of solver calls 	<ul style="list-style-type: none"> § very robust, can not diverge § easy to apply 	<ul style="list-style-type: none"> § good for problems with many local minimas 	<ul style="list-style-type: none"> § very effective, particularly SRSM § trade-off studies on RS § filter out noise, smoothing of results
	<ul style="list-style-type: none"> § can diverge § can stuck in local minimas § step-size dilemma for numerical gradients 	<ul style="list-style-type: none"> § bad convergence, not effective § Chooses best observation – may not be representative of a good (robust) design 	<ul style="list-style-type: none"> § many solver calls, only suitable for fast solver runs § Chooses best observation – may not be representative of a good (robust) design 	<ul style="list-style-type: none"> § approximation error § verification run might be infeasible § number of variables control minimum number of required runs

Stochastic Analysis - Goals

Statistical Quantities of Output (Response) due to Variation of Input (Parameter)

- *Mean*
- *Standard deviation*
- *Distribution function*

Significance of Parameter with respect to Responses

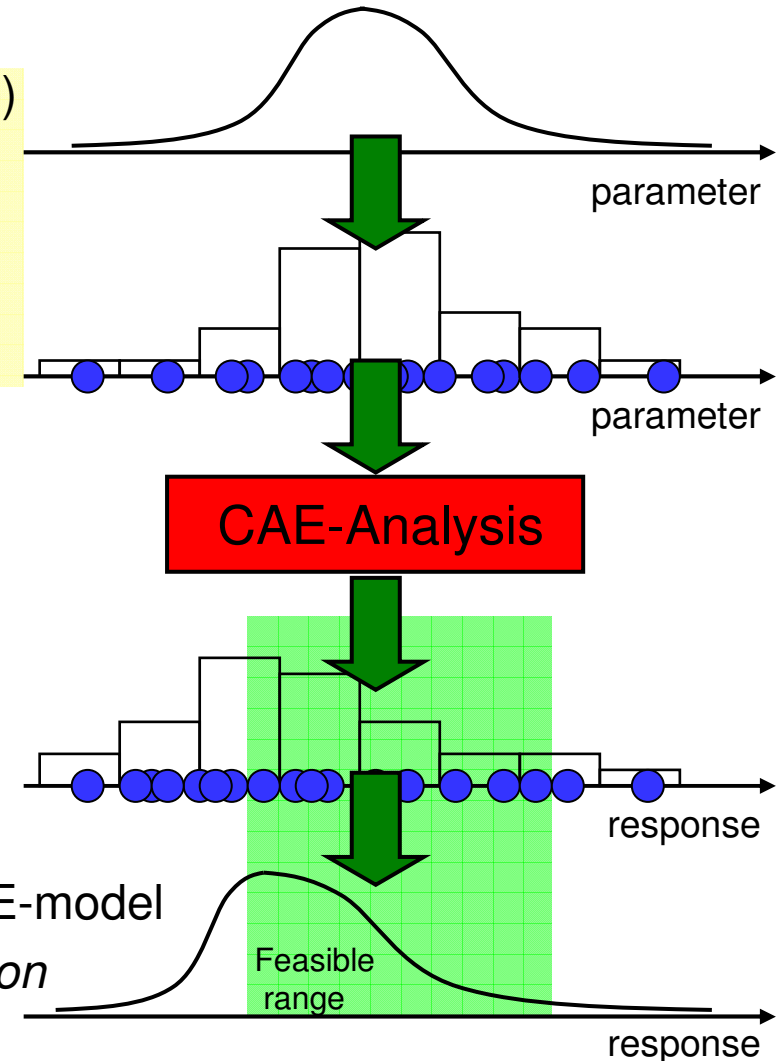
- *Correlation analysis*
- *Stochastic contributions*
- *ANOVA – analysis of variance*

Reliability Issues

- *Probability of failure*

Visualization of statistical quantities on FE-model

- *Spatial detection of variation/correlation*



Methodologies – Robustness Investigations

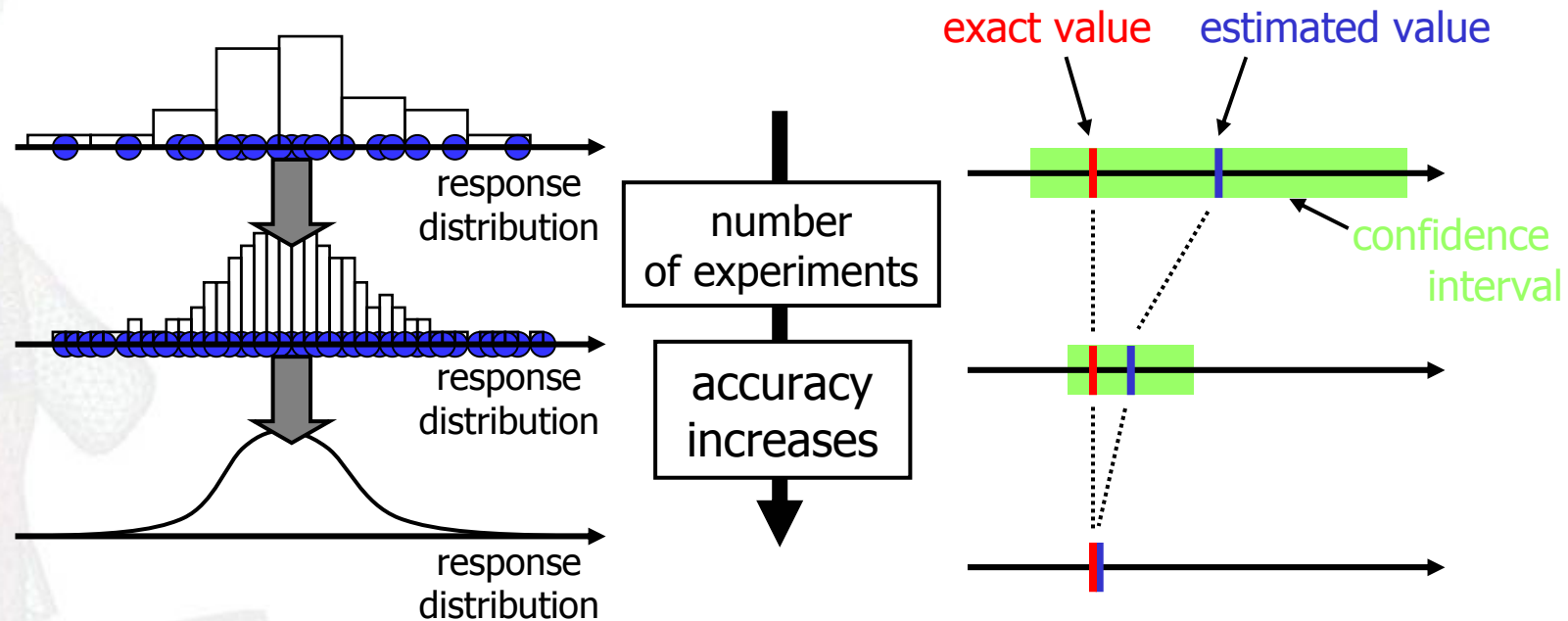
- Introduction/Features
- Methods – Optimization
- Methods - Robustness**
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- Version 3.2 / Outlook

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⌚ Statistical Quantities of Output due to Variation of Input

■ Direct Monte Carlo Sampling

- *Latin Hypercube sampling*
- *Large number of FE runs (100+)*
- *Consideration of confidence intervals for mean, std. dev., correlation coeff.*



Methodologies – Robustness Investigations

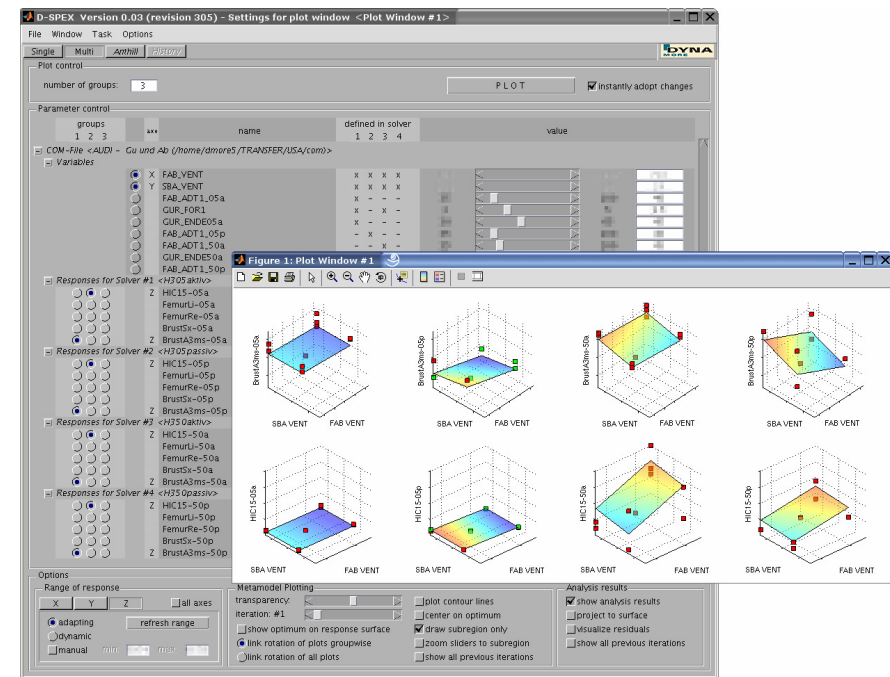
- Introduction/Features
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Statistical Quantities of Output due to Variation of Input

Monte Carlo using Meta-Models

- Response Surface / Neural Network
- Medium number of FE runs (10 – 30+)
- Number of runs depend on the dimension of the problem (number of variables) and the type of the response surface
- Identify design variable contributions clearly
- Exploration of parameter space
->D-SPEX



Multi Meta-Model exploration with D-SPEX

Methodologies – Robustness Investigations

- Introduction/Features
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Significance of Parameter with respect to Responses

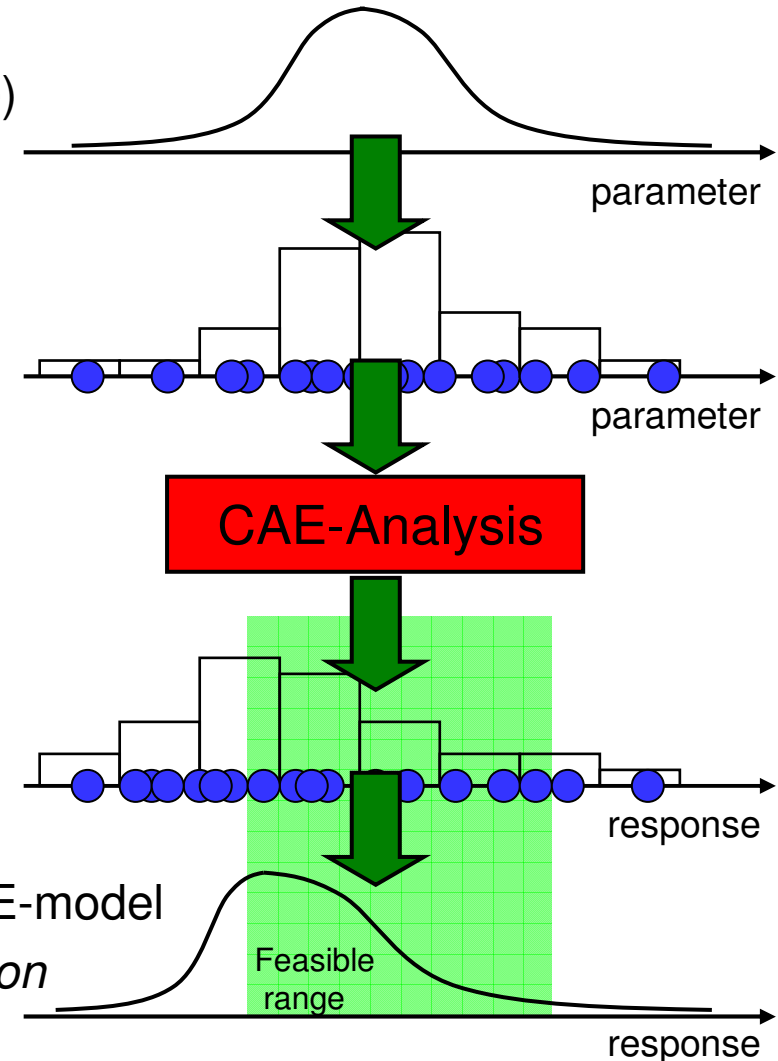
- correlation analysis
- stochastic contributions
- ANOVA – analysis of variance

Reliability Issues

- Probability of failure

Visualization of statistical quantities on FE-model

- Spatial detection of variation/correlation



Methodologies – Robustness Investigations

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Stochastic Analysis - Goals

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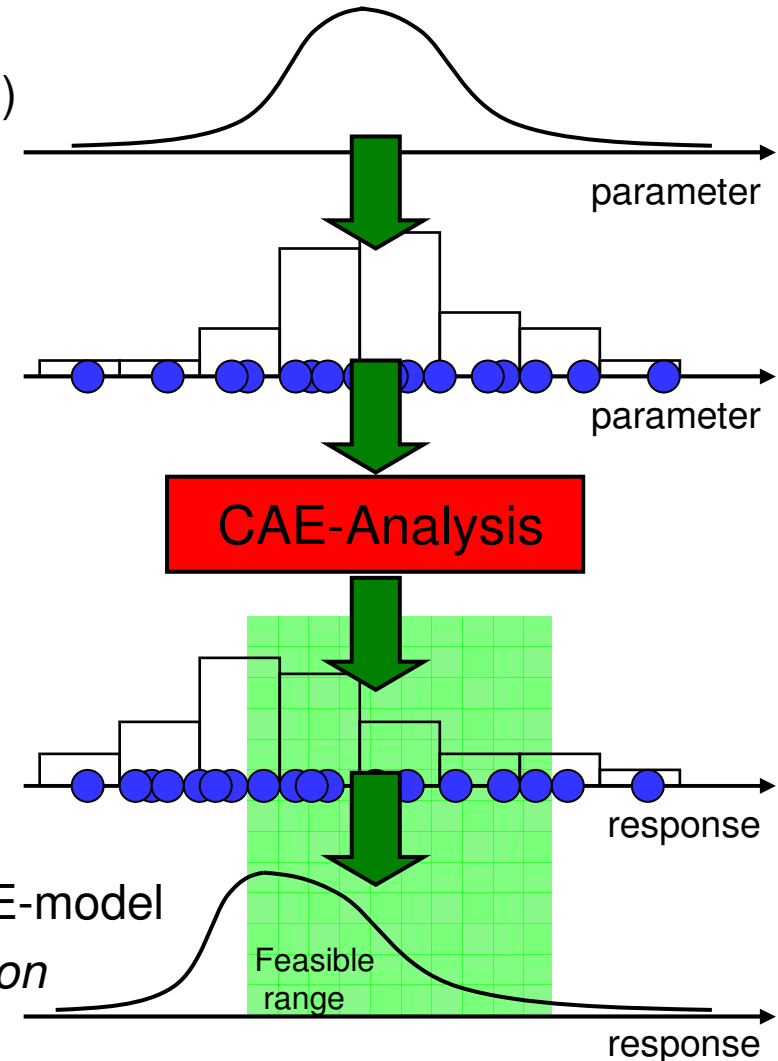
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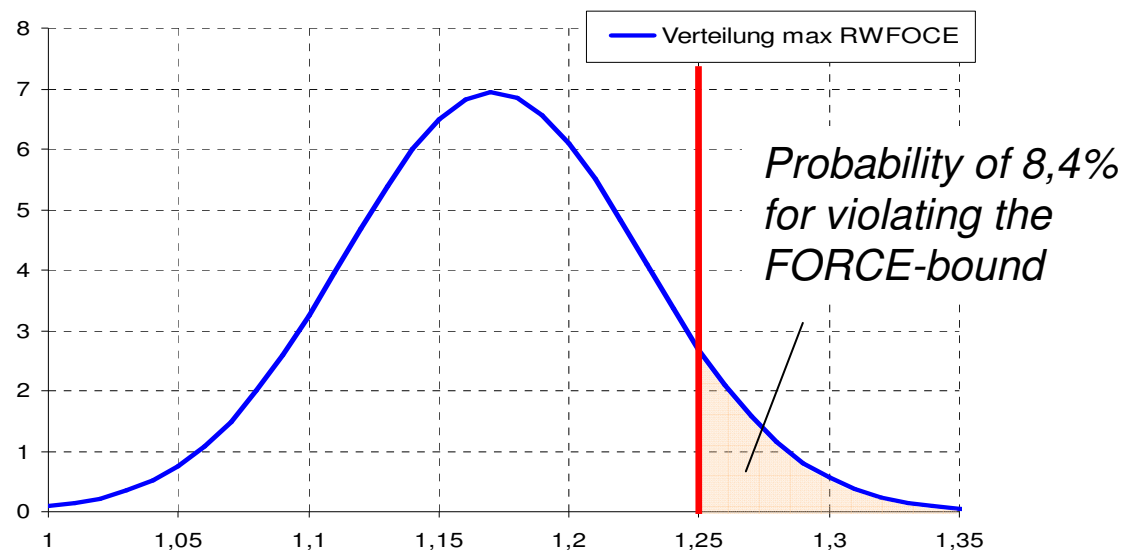
Methodologies – Robustness Investigations

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⌘ Reliability Analysis

- n Probability of failure
- n Evaluation of confidence interval
- n Prediction error (confidence interval) depends
 - n *on the number of runs*
 - n *on the probability of event*
 - n *not on the dimension of the problem (number of design variables)*



Methodologies – Robustness Investigations

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Stochastic Analysis - Goals

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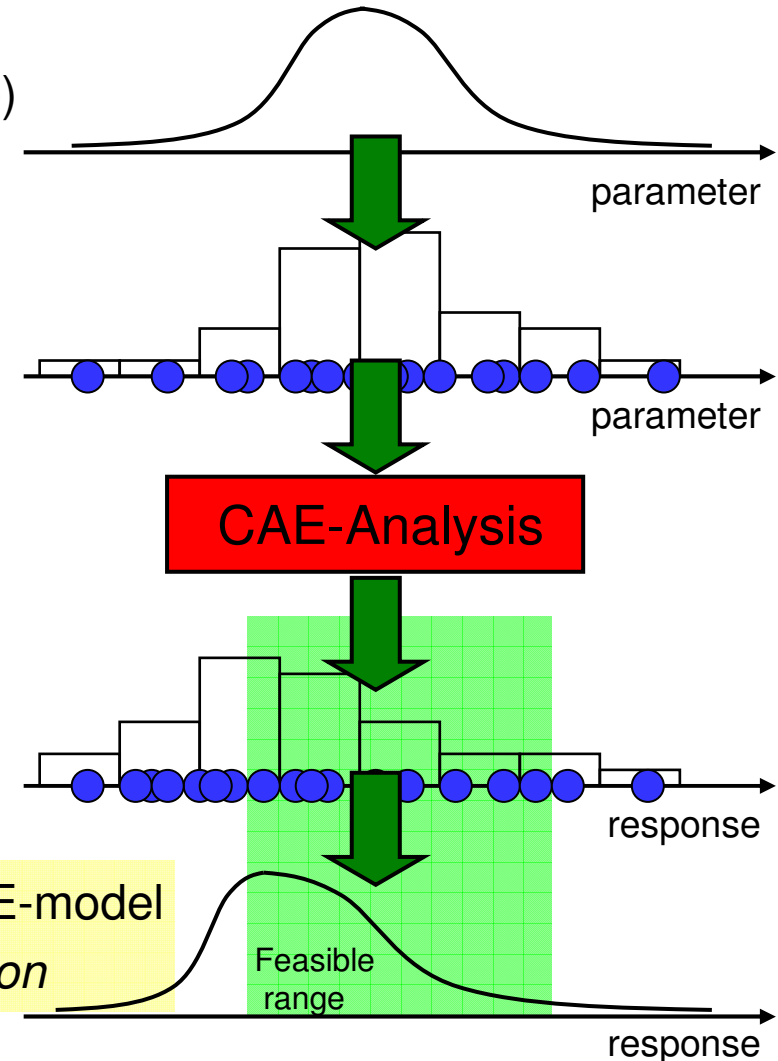
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- *Probability of failure*

Visualization of statistical quantities on FE-model

- *Spatial detection of variation/correlation*



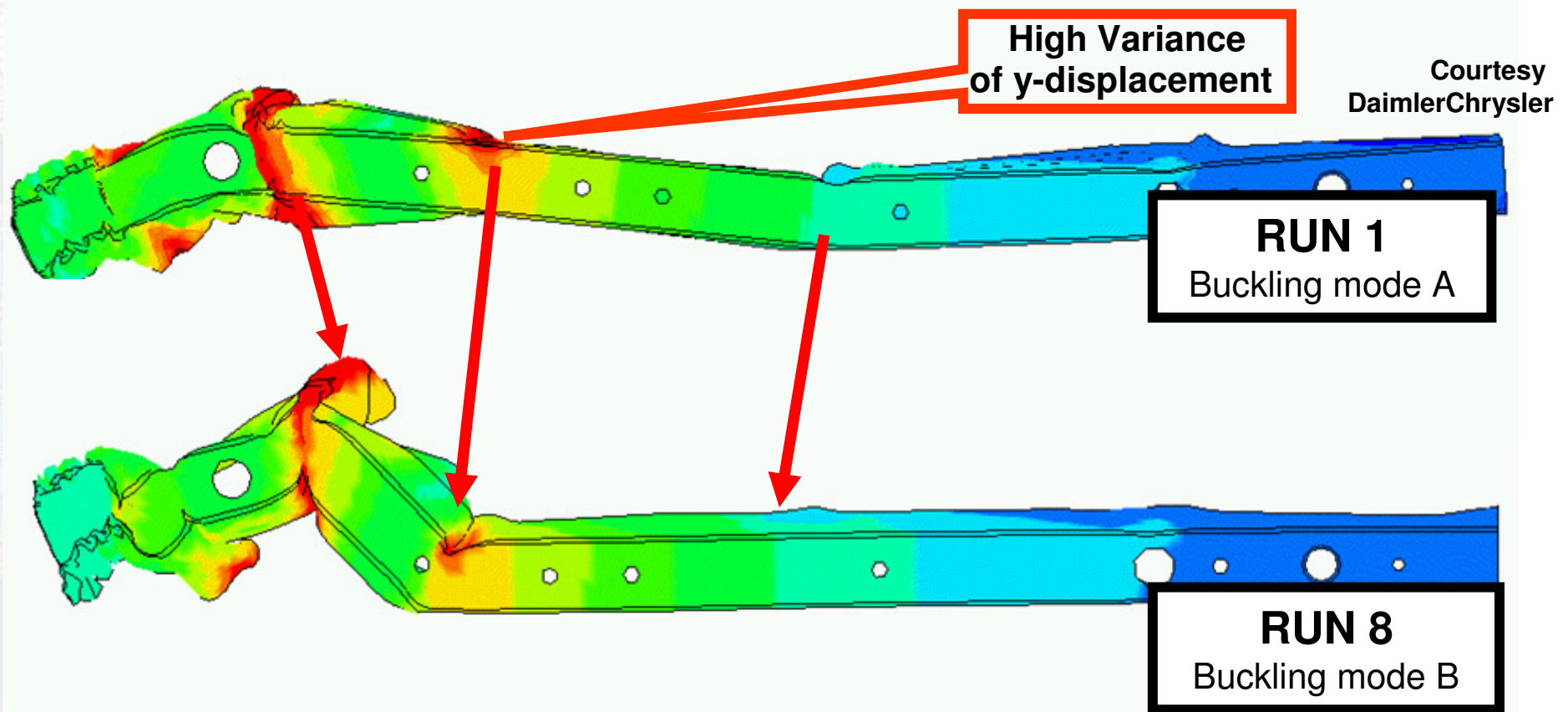
Methodologies – Robustness Investigations

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Visualization of Statistical Quantities on FE-model

- Standard deviation of y-displacements of each node (40 runs)



Example I - Optimization

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Optimization of an Adaptive Restraint System

- Four Different Front-Crash Load Cases (FMVSS 208)

Dummy	56 km/h – belted	40 km/h – not belted
Hybrid III 5th Female	H305a (ktiv)	H305p (assiv)
Hybrid III 50th Male	H350a (ktiv)	H350p (assiv)

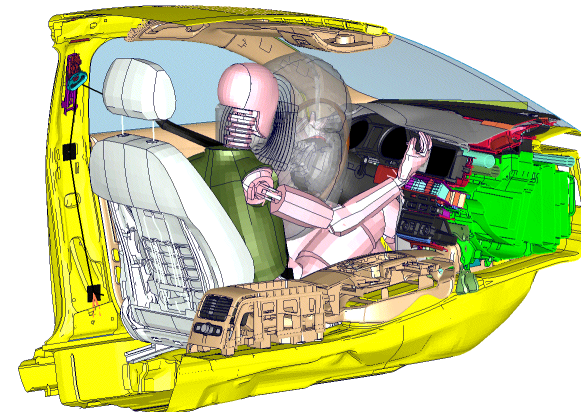
- PAM-Crash Model

- about 500000 elements
- wall clock simulation time ~19 h,
4 cpus, distributed memory

- Load Case Detection available

- Differentiation of the loadcases
belted / not belted and
“Hybrid III 5th Female“ / „Hybrid III 50th Male“ possible

- Trigger time for seatbelt, airbag and steering column might be different



Example I - Optimization

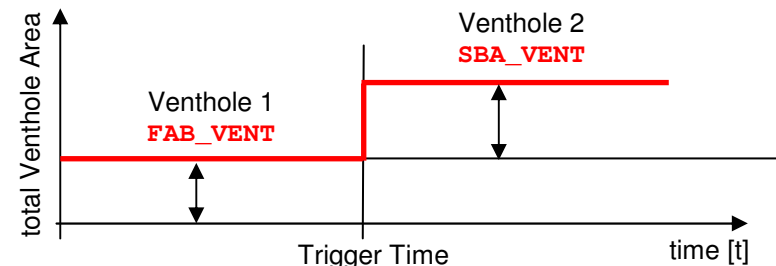
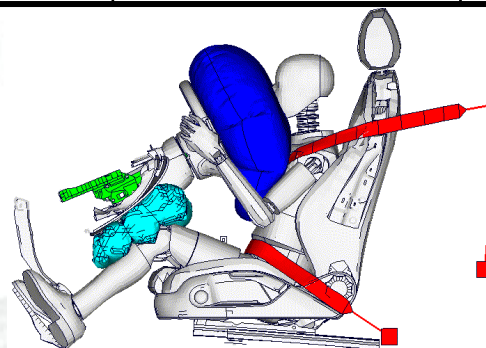
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Design Variables

Adaptive Airbag Deployment (6 Variables)

	H305a 5%-dummy, belted	H305p 5%-dummy, not belted	H350a 50%-dummy, belted	H350p 50%-dummy, not belted
Area Venthole1 Lower – Upper B.	FAB_VENT	FAB_VENT	FAB_VENT	FAB_VENT
Area Venthole2 Lower – Upper B.	SBA_VENT	SBA_VENT	SBA_VENT	SBA_VENT
Trigger Time Lower – Upper B.	FAB_ADT1_05a	FAB_ADT1_05p	FAB_ADT1_50a	FAB_ADT1_50p



- FAB_ADT1_05a**
- FAB_ADT1_50a**
- FAB_ADT1_05p**
- FAB_ADT1_50p**

Example I - Optimization

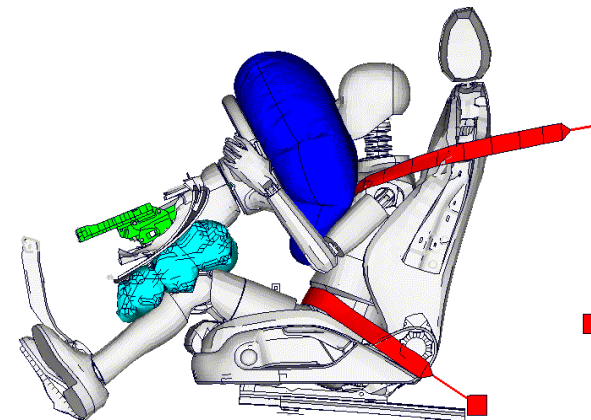
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Design Variables

Adaptive Steering Column (5 Variables)

	H305a 5%-dummy, belted	H305p 5%-dummy, not belted	H350a 50%-dummy, belted	H350p 50%-dummy, not belted
Force Level StCo	LKS_SKAL	LKS_SKAL	LKS_SKAL	LKS_SKAL
Lower – Upper Bound				
Trigger Time	LKS_DOWN05a	LKS_DOWN50a	LKS_DOWN05p	LKS_DOWN50p
Lower – Upper Bound				



Example I - Optimization

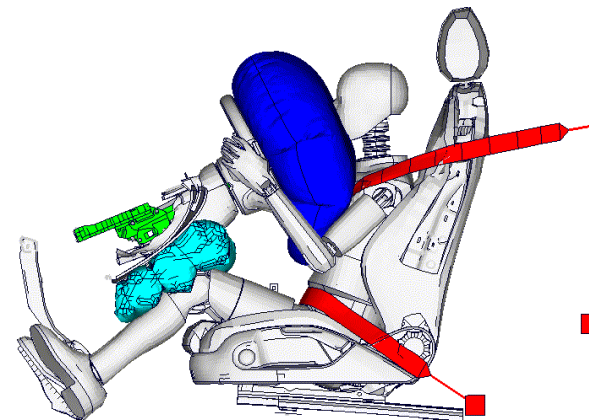
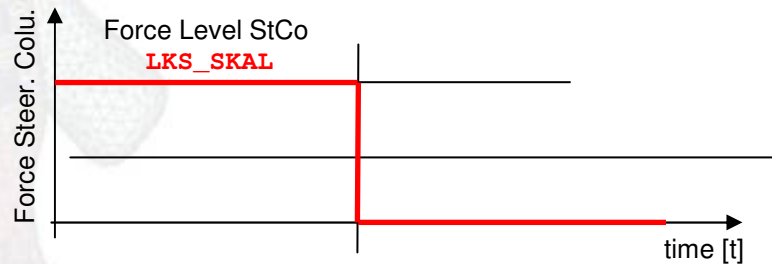
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Design Variables

Adaptive Steering Column (5 Variables)

	H305a 5%-dummy, belted	H305p 5%-dummy, not belted	H350a 50%-dummy, belted	H350p 50%-dummy, not belted
Force Level StCo	LKS_SKAL	LKS_SKAL	LKS_SKAL	LKS_SKAL
Lower – Upper Bound				
Trigger Time	LKS_DOWN05a	LKS_DOWN50a	LKS_DOWN05p	LKS_DOWN50p
Lower – Upper Bound				



Example I - Optimization

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Optimization Problem

Objective

Minimize Thorax Acceleration

- > min BrustA3ms-05a
- > min BrustA3ms-50a
- > min BrustA3ms-05p
- > min BrustA3ms-50p

Constraints < 80% of regulation requirements

Head Injury Coefficient (15ms)

- > HIC15-05a
- > HIC15-50a
- > HIC15-05p
- > HIC15-50p

Femur Forces (left/right)

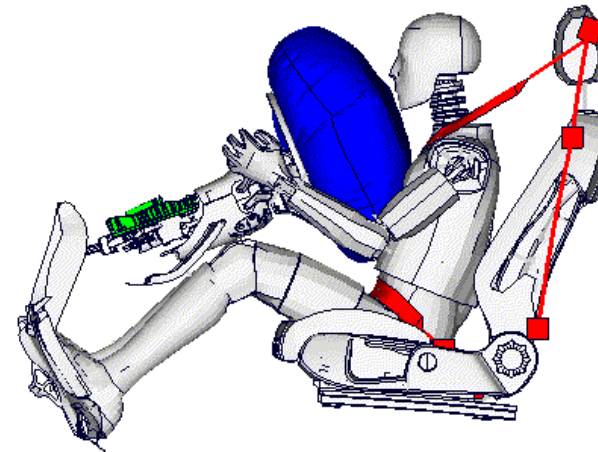
- > FemurLi-05a
- > FemurLi-50a
- > FemurLi-05p
- > FemurLi-50p

Thorax Intrusion

- > BrustSx-05a
- > BrustSx-50a
- > BrustSx-05p
- > BrustSx-50p

Thorax Acceleration

- > BrustA3ms-05a
- > BrustA3ms-50a
- > BrustA3ms-05p
- > BrustA3ms-50p



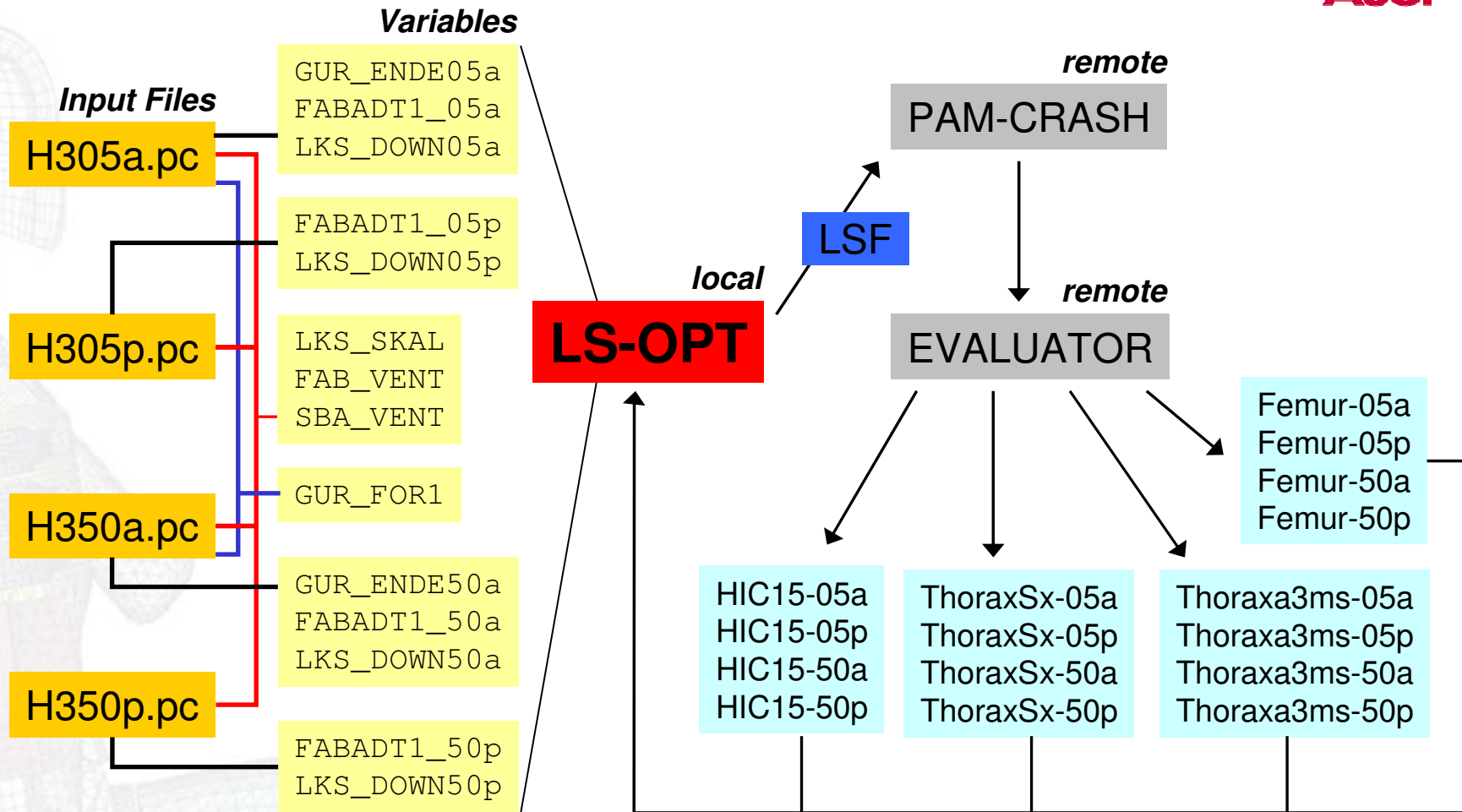
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Process Work Flow



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Application of Optimization

■ Preferred Configuration at AUDI

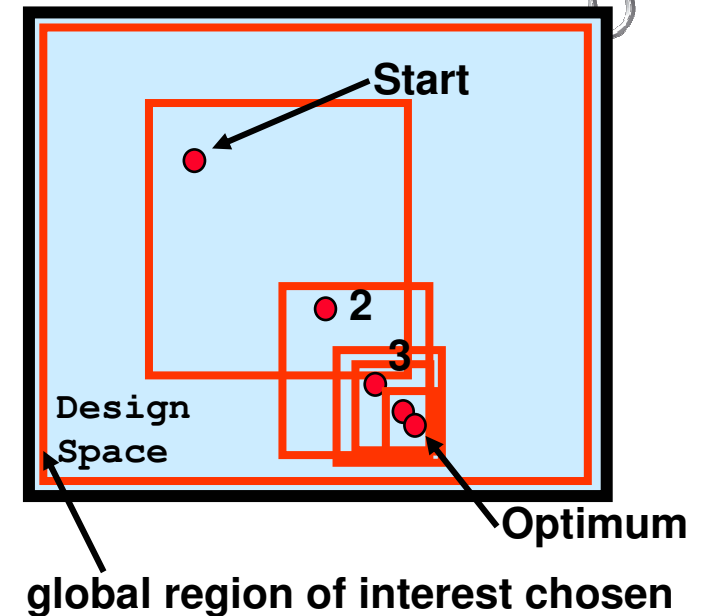
- *Adaptive Restraint System only for Airbag and Seatbelt*
- *Reduction to 9 Variables in total (active=6, passive=3)*

■ LS-OPT Approach: Successive Response Surface Methodology (SRSM) using **linear** polynomial approximations

- *34 runs per iteration*
- *D-optimal Design of Experiments (DOE)*

■ Results

- *8 iterations - total runs: 276*
- *all constraints are fulfilled*
- *minimization of multi-objective (second step) not applied*

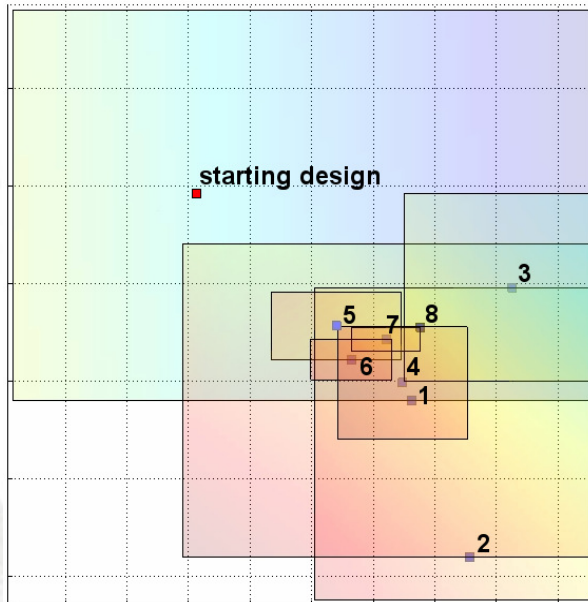


Example I - Optimization

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- Examples - Robustness
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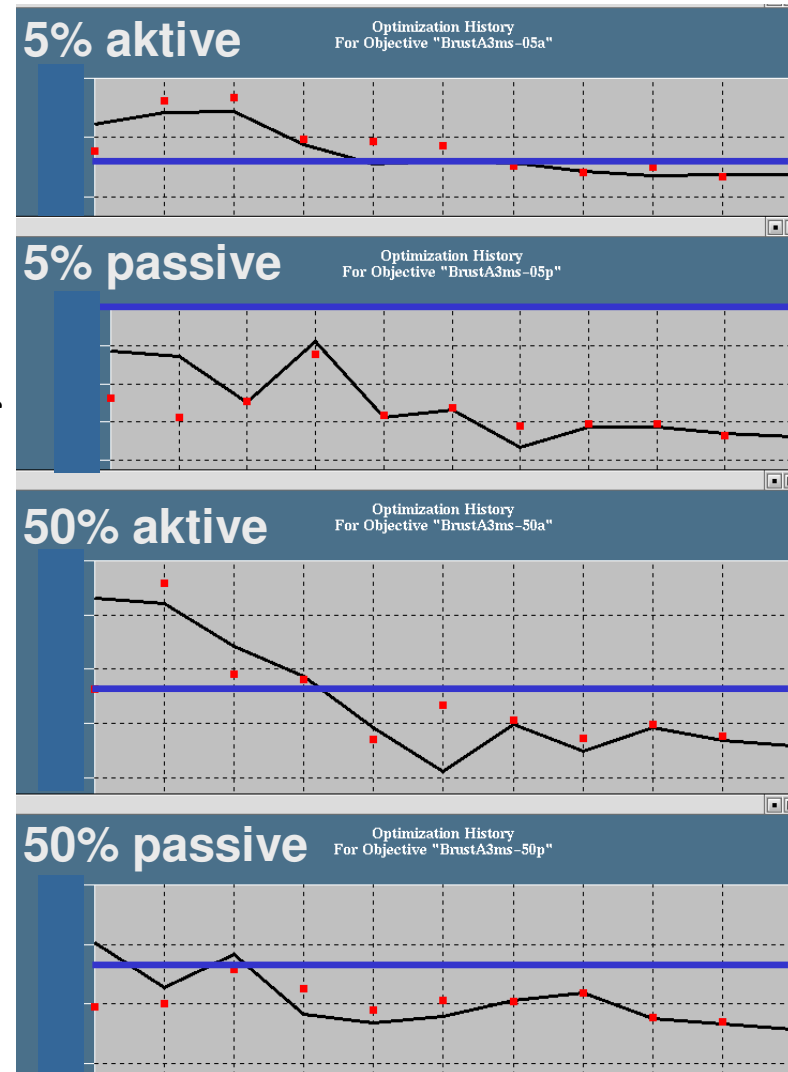


Optimization Progress



a result which meets all requirements is gained in 8 iterations, each with 34 shots

History of Thorax Acceleration



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⌚ Parameter Identification of Plastic Material

- Material properties: nonlinear visco-elastic behaviour
- LS-DYNA hyperelastic/viscoelastic formulation - *MAT_OGDEN_RUBBER (#77)
- Hyperelasticity

$$W = \sum_{i=1}^3 \sum_{j=1}^n \frac{\mu_j}{\alpha_j} (\lambda_i^{\alpha_j} - 1) + \frac{1}{2} K (J - 1)^2$$

- Prony series representing the viscos-elastic part (Maxwell elements):

$$g(t) = \sum_{m=1}^N G_m e^{-\beta_m t} \quad ; \quad N=1, 2, 3, 4, 5, 6 \quad ; \quad \sigma_{ij} = \int_0^t g_{ijkl}(t - \tau) \frac{\partial \varepsilon_{kl}}{\partial \tau} d\tau$$

Example I - Optimization

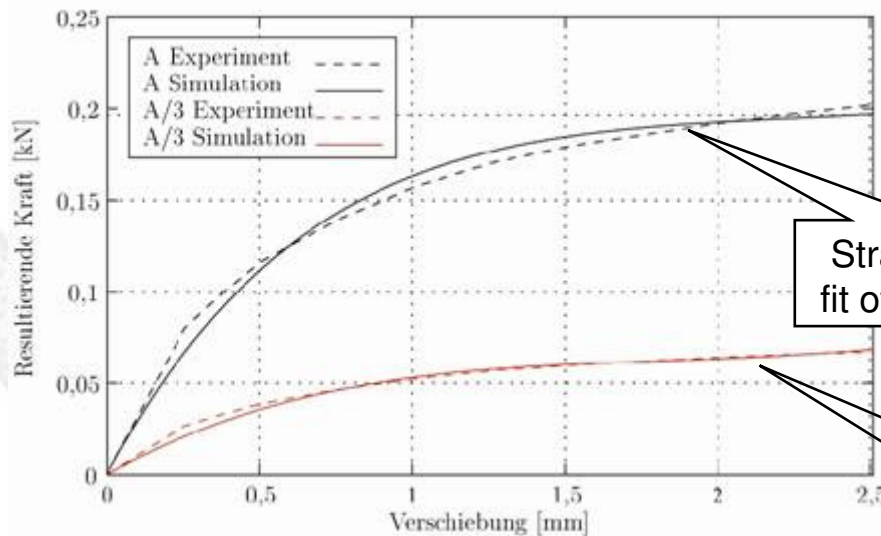
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Parameter Identification of Plastic Material

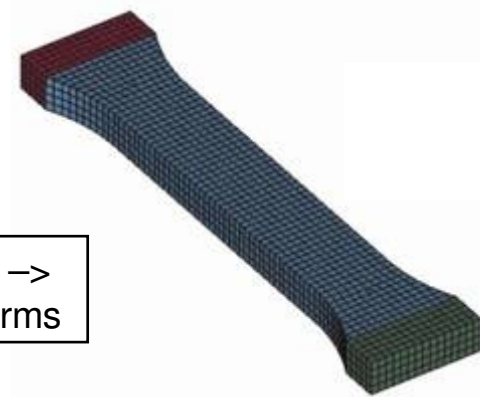
- Minimize the distance between experimental curve and simulation curve
- Least-Squares Objective Function

$$F(\mathbf{x}) = \sum_{p=1}^P \{ [y(\mathbf{x}) - f(\mathbf{x})]^2 \} \rightarrow \min F(\mathbf{x})$$



Strain rate A →
fit of Prony terms

quasi-static curve –
> Ogden fit



Example III – Optimization

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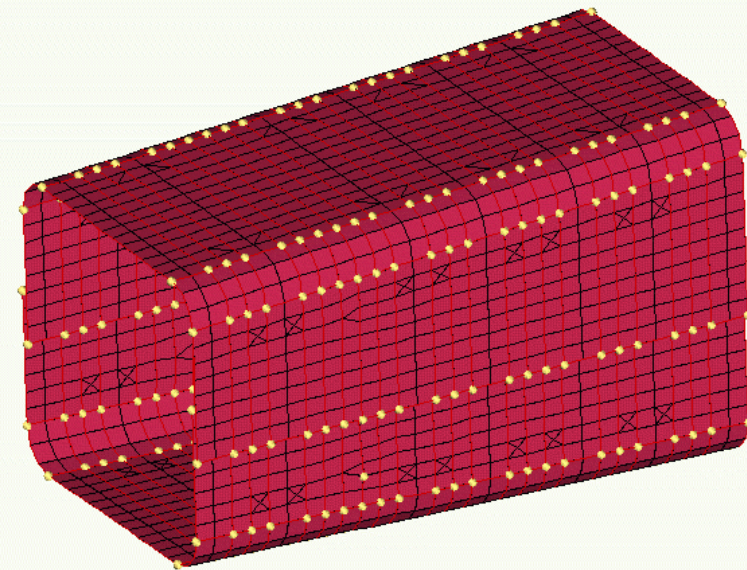
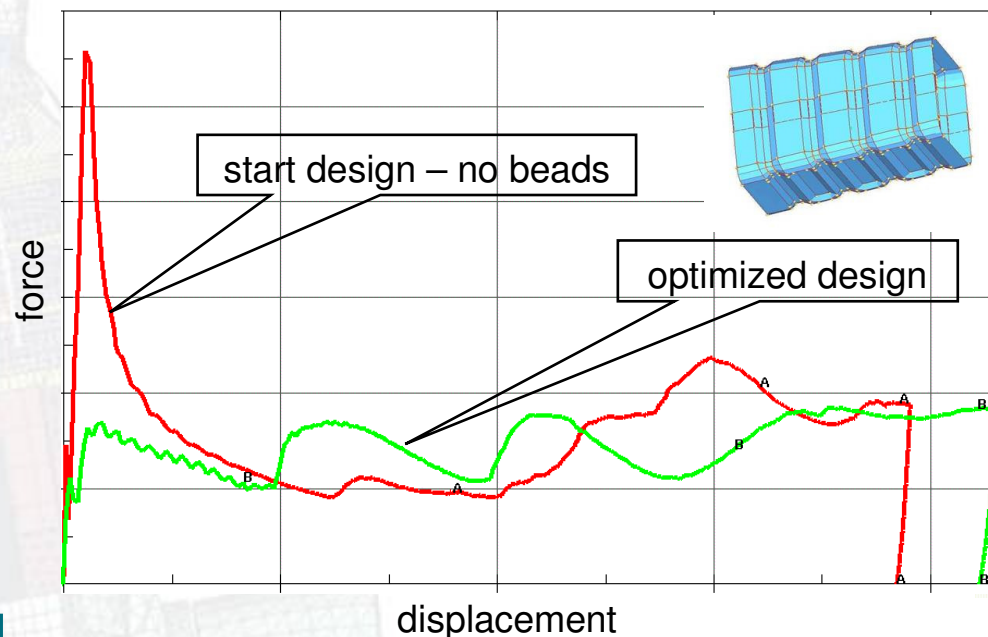


Shape Optimization of a Crash Box

Scope of optimization:

- minimize the maximum crash force
- steady-going force progression

Shape variation by using Hypermorph and LS-OPT (20 design variables)



Example I – Robustness

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⌘ Robustness Investigations – Monte Carlo Analysis

n Variation of sheet thicknesses and yield stress of significant parts in order to consider uncertainties

n Normal distribution is assumed

n T_{1134} (Longitudinal Member) $mean = 2.5mm; \sigma = 0.05mm$

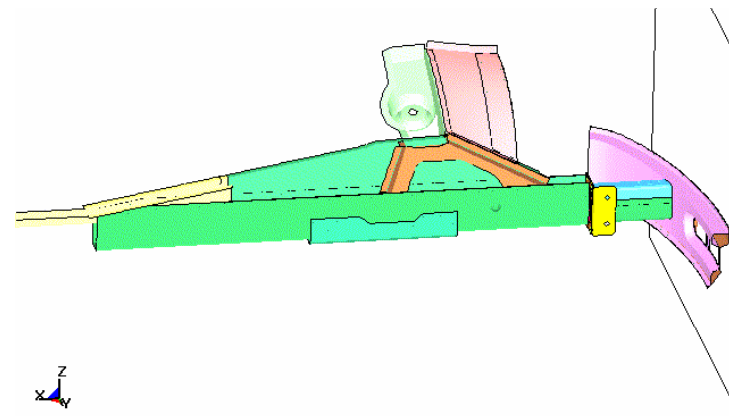
n T_{1139} (Closing Panel) $mean = 2.4mm; \sigma = 0.05mm$

n T_{1210} (Absorbing Box) $mean = 0.8mm; \sigma = 0.05mm$

n T_{1221} (Absorbing Box) $mean = 1.0mm; \sigma = 0.05mm$

n SF_{1134} (Longitudinal Member) $mean = 1.0 ; \sigma = 0.05$

n Monte Carlo analysis using 182 points (Latin Hypercube)



Example I – Robustness

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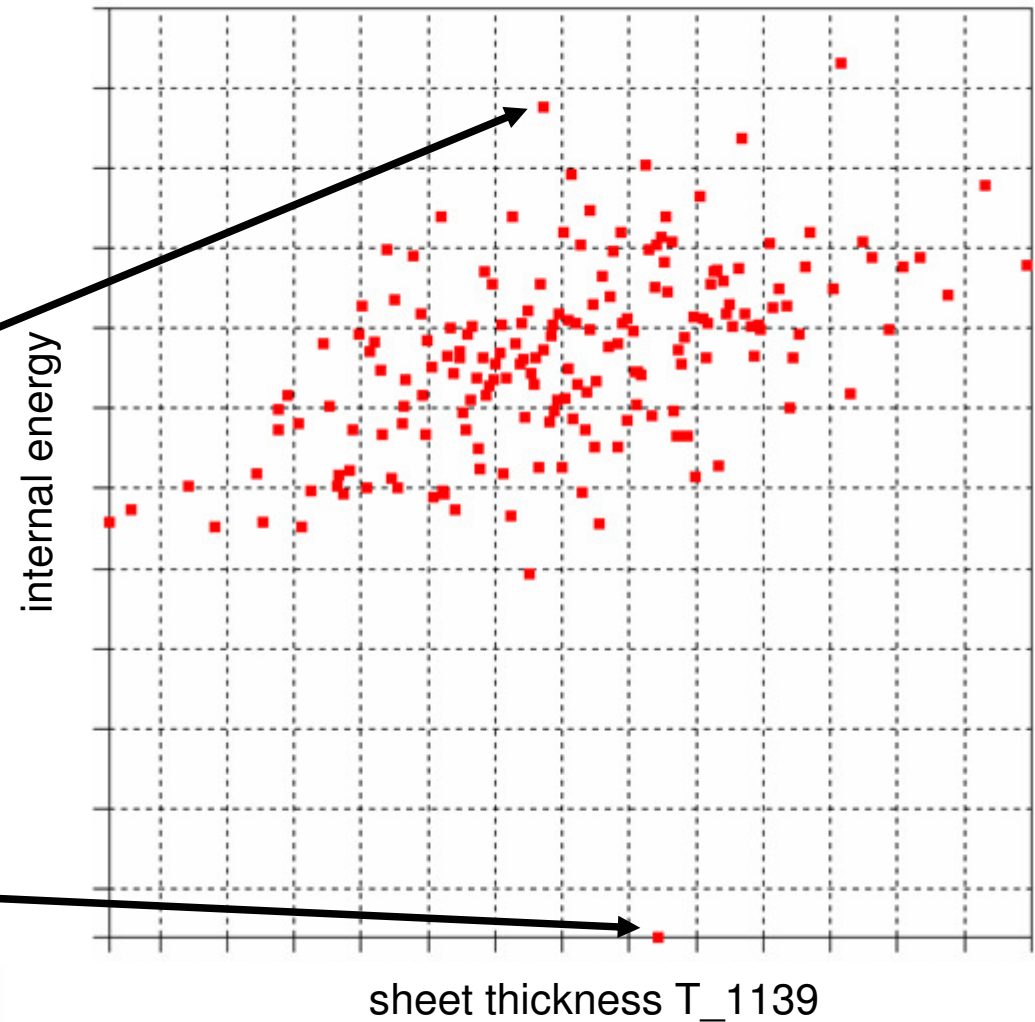
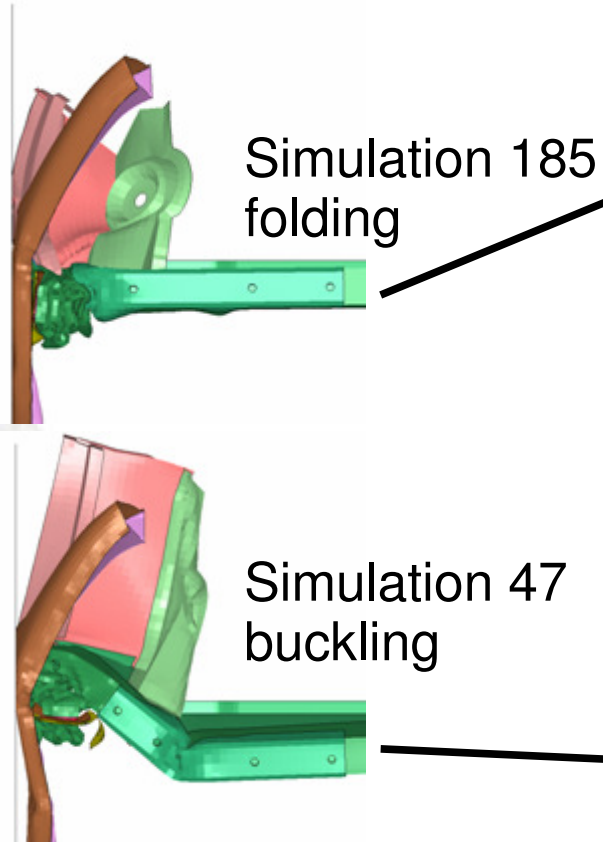


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Tradeoff Plot

Monte Carlo Simulation

Identification of Clustering



Example I – Robustness

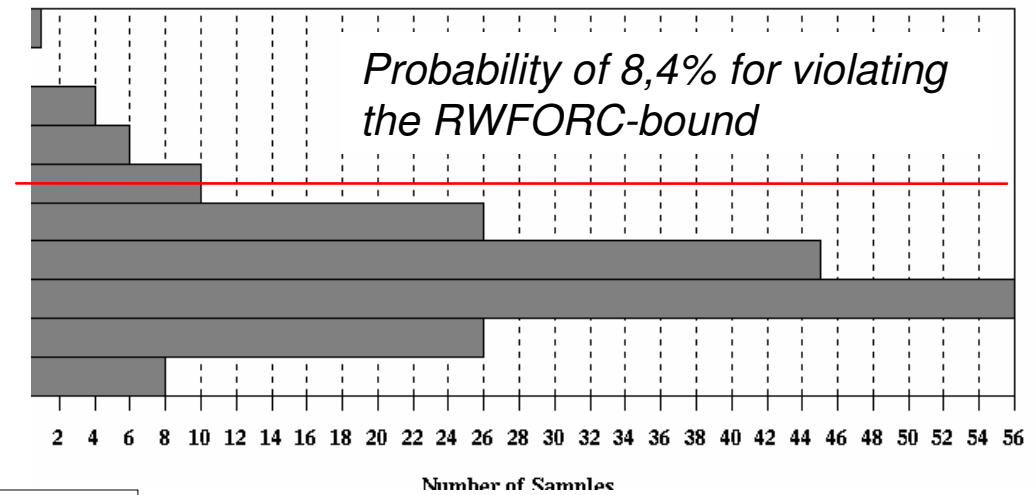
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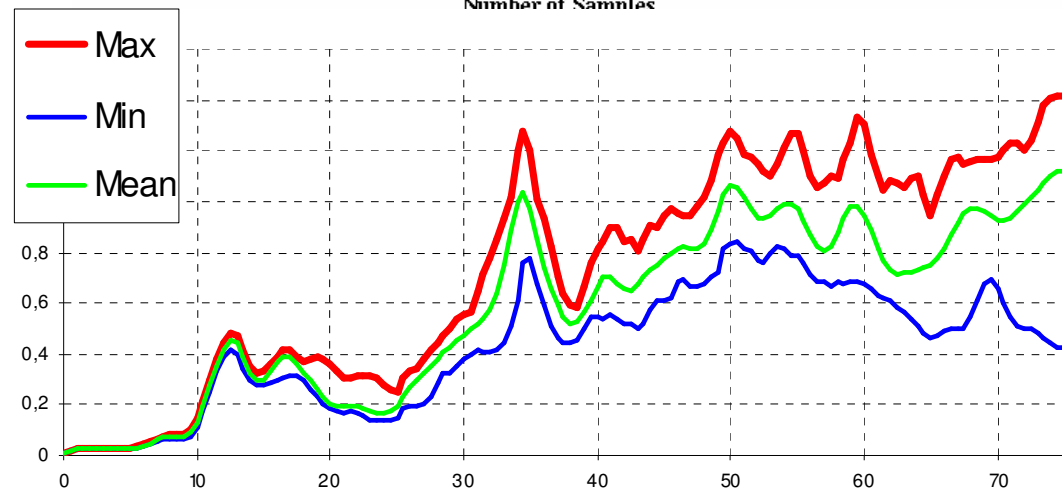
Reliability Analysis

- Histogram of distribution
- Probability of exceeding a constraint-bound



Min-Max Curves

- Plot of minimum, maximum and mean history values
- Gives a confidence interval of history values



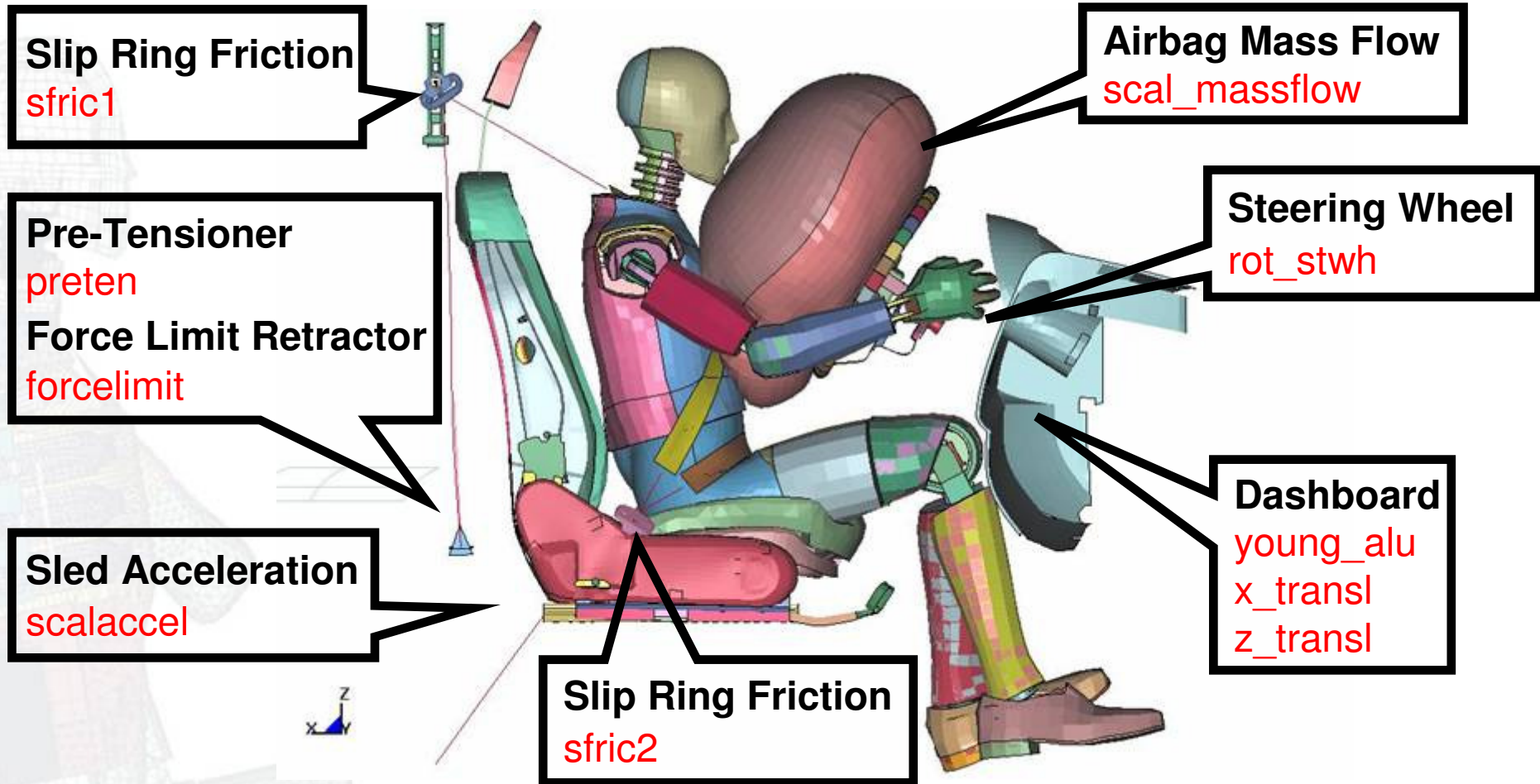
Example II – Robustness

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Design Variables - Uncertainties in Test Set-Up



Example II – Robustness

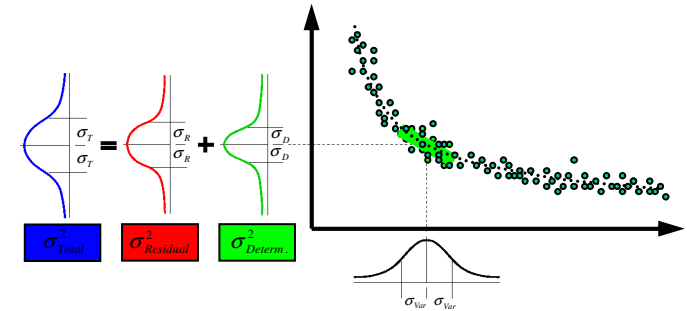
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Stochastic Contribution - Results of 30 Experiments

Design Variable	Standard Deviation of Design Variable	Standard Deviation Contribution					
		HIC36	max_chest_intru	max_b_f_shoulder	max_bf_pelvis	max_chest	max_pelvis
scalaccel	2,5%	3,1%	1,5%	0,1%	2,3%	1,9%	2,9%
sfri1	25,0%	1,3%	0,6%	4,1%	1,8%	0,7%	0,7%
sfri2	25,0%	0,5%	0,6%	0,1%	3,7%	0,1%	0,1%
preten	4,4%	0,0%	0,5%	0,0%	1,1%	0,3%	0,2%
forcelimit	5,6%	1,3%	0,4%	4,4%	0,6%	1,4%	0,2%
rot_stwh	4,8%	0,5%	0,1%	0,1%	0,0%	0,1%	0,1%
transl_x	50,0%	0,1%	0,1%	0,7%	4,5%	0,5%	0,8%
transl_z	50,0%	1,2%	1,0%	0,3%	1,6%	0,2%	0,9%
scalmassflow	5,0%	1,8%	1,8%	0,6%	2,2%	0,6%	0,9%
young_alu	5,0%	0,3%	0,3%	0,0%	0,5%	0,1%	0,1%
all variables		4,3%	2,8%	6,1%	7,2%	2,6%	3,4%
residuals		4,7%	1,9%	1,8%	6,0%	3,5%	2,3%
Total		6,4%	3,4%	6,3%	9,4%	4,3%	4,1%



Contribution of variation of design variables to variation of results

Meta-model space

Residual space

Total Variation

Example II – Robustness

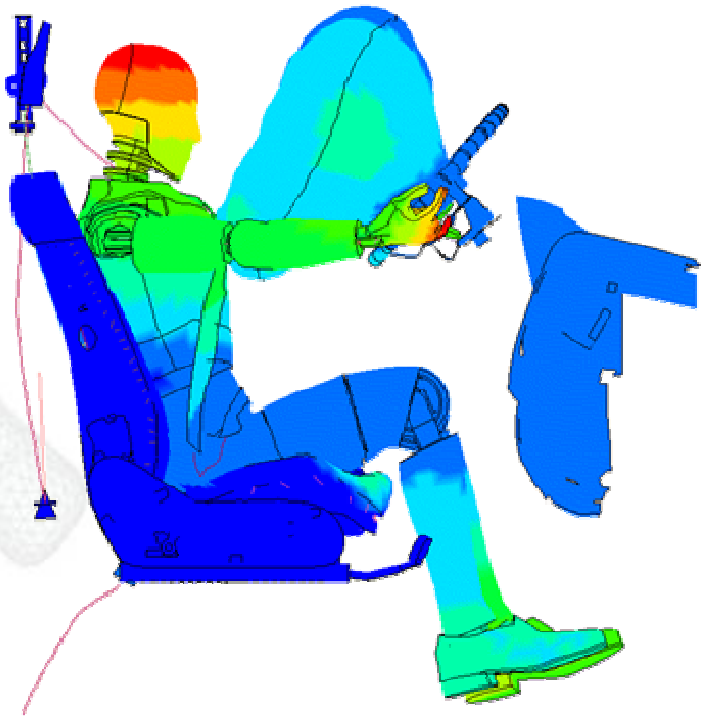
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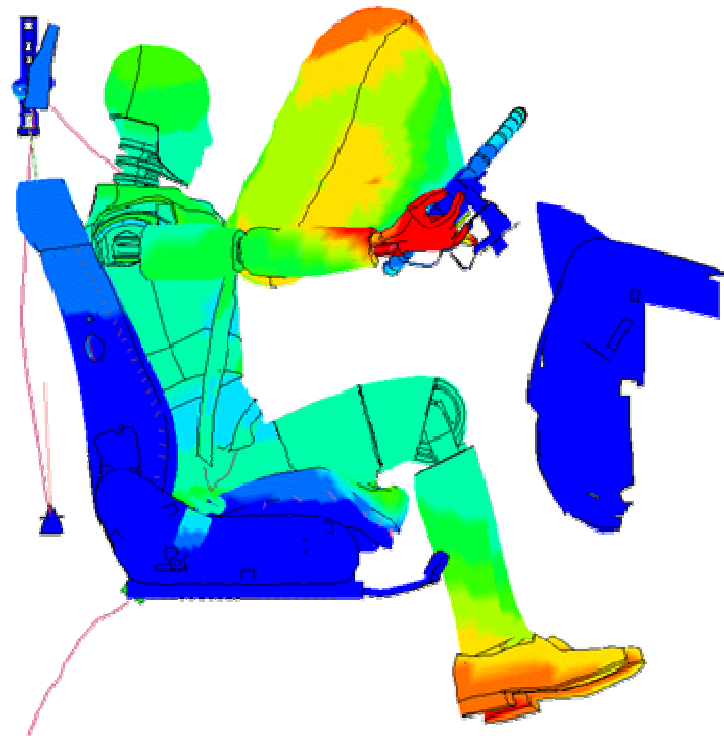
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Standard deviation of x-displacements of each node (120 runs)

(a) Deterministic (Meta-Model)



(b) Residual (Outliers)



⌘ Version 3.3

■ Improvements of Meta-Models

- *Implementation of Radial Basis Functions*
- *Speed/Performance enhancements of Neural Networks*

■ Genetic Algorithm (MOGA – NSGA-II)

- *Improve of Multi-Objective Optimization (Pareto Fronts)*
- *Direct GA available*

■ Tied ANSA Interface

- *User friendly coupling of ANSA*

■ Extra Input Files

- *Additional Input Files containing Variables can be specified*
- *For other solvers than LS-DYNA*

⚙ Version 3.3

■ DYNASTATS for Metal Forming

- *Available for adaptive meshing*
- *Mapping of nodal/element results onto reference mesh*

■ 3-D metamodel plot enhancements

- *Activate Post-Processor on point selection*
- *Add value list display on point selection (similar to 2D)*

■ ANOVA chart enhancements

- *Positive/negative correlation*

■ DOE-Task for Sensitivities and Variable Screening

- *Dedicated task for DOE-Study (no optimization)*

Version 3.3/Outlook

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⌘ **Version 3.3**

- Interface for User-defined Meta Models
- Summary Report File
- Import of Check Points
 - *Calculation of predicted values for user-defined points*

↳ Outlook

- Generic File extractor
 - *Extraction of values from any ASCII input file*
- Visualization of “Pareto Fronts” for Multi-Objective Optimization (MOO)
 - *Difficult for more than 3 objectives*
- *Correlated Input Variables for Stochastic Investigations*
- Additional injury criteria (DYNA extraction)
 - *IIHS, neck/tibia indices,...*



Thanks for your attention!

