EADS INNOVATION WORKS

VORKS

2nd International Workshop on Validation of Computational Mechanics Models

Uncertainty Quantification and Robust Design Optimisation

5th November 2013

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Uncertainties of Real Systems

Scatter of geometry

- Wall thickness
- Tolerances in shape or measure

Scatter of material properties

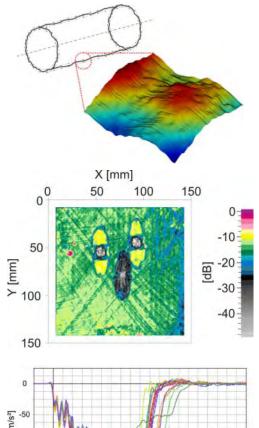
- Density
- Young's modulus
- Fibre orientation for composite materials

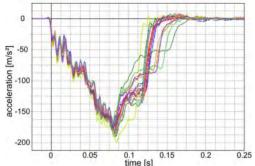
Scatter of loads

Acceleration impulse

Scatter of initial and boundary conditions

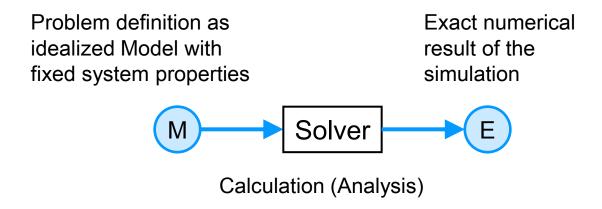
- Initial velocity
- Spring and damper characteristics







Deterministic Simulation



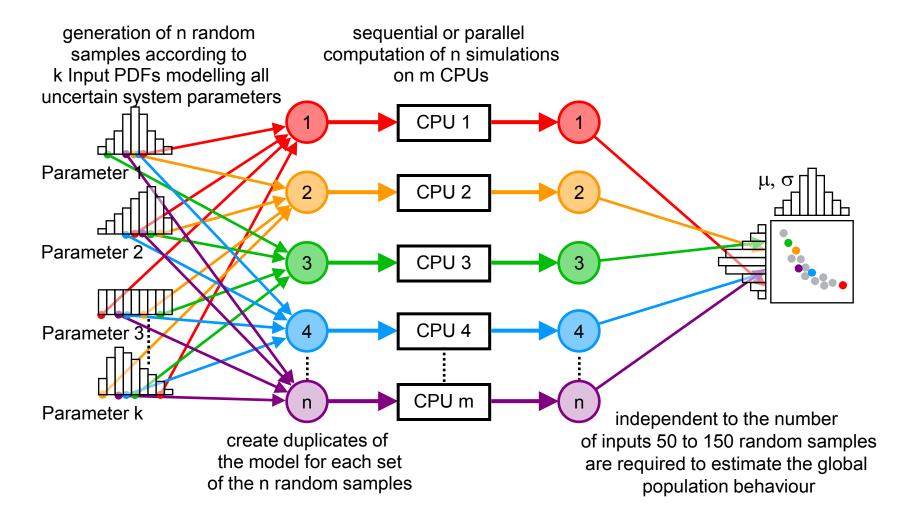
Fixed link between Model and numerical analysis result

Disadvantages:

- Cause-effect relationships of the system may remain concealed
- Influence of scatter due to production and usage tolerances not quantifiable
- No information about system robustness

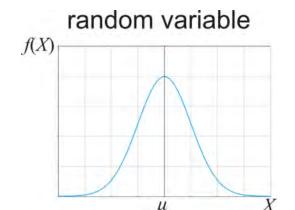


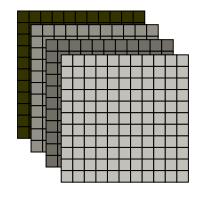
Stochastic Simulation (Monte Carlo Method)



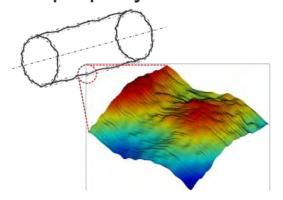


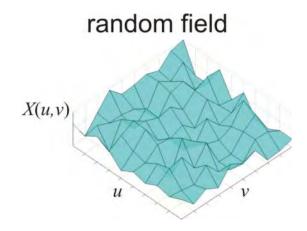
Mathematical Models of Uncertainties





real property distribution



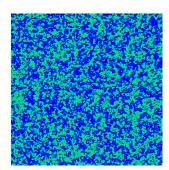




Some Random Field Properties

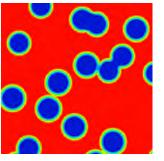
Description by a finite set of parameters is needed

Gaussian random field: mean and covariance function



Structural features like shape or connectivity:

- Short-range correlation model
- 2nd-order statistics: lowly-structured morphologies
- 3rd-order statistics: medium-structured morphologies



Random fields may have highly complex structure

- Composites with various fibre orientations, cavities,...
- Efficient modelling method is required (e.g. MCS)



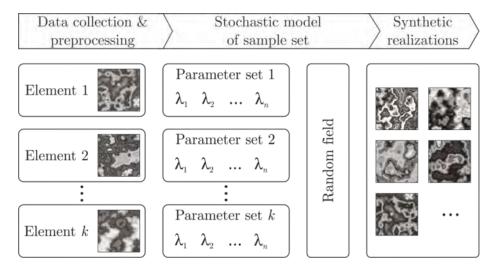


Pattern Recognition Based Parameterisation

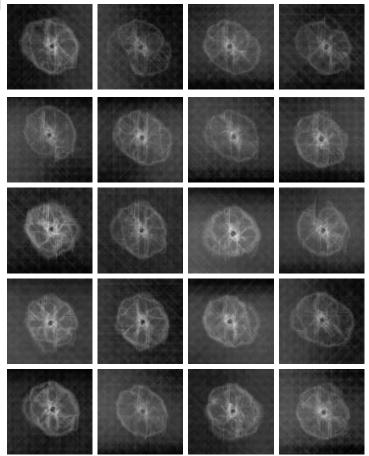
Random field: n-dimensional image

Approach: Methods derived from image processing

Objective: Pattern recognition based parameterisation method capable to represent highly structured morphology



Application example: 20 X-ray scans of damaged CFRP-plates

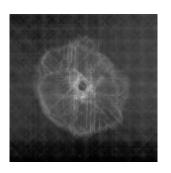


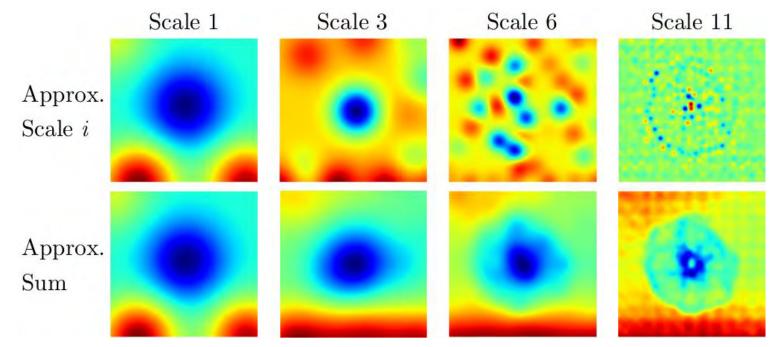


Feature Extraction and Multiscalar Approximation

Basic idea: split discrete signal *x* into limited frequency bands

- Local extreme is called Blob
- Each Blob is approximated by one Gaussian kernel
- Gaussian kernels positioned at extremes of Diff.







Probabilistic Model for Synthetic Realizations

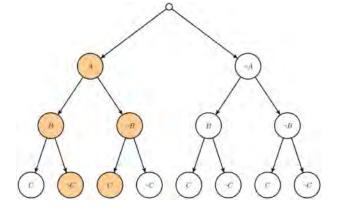
Probabilistic Model by means of kernel density estimation

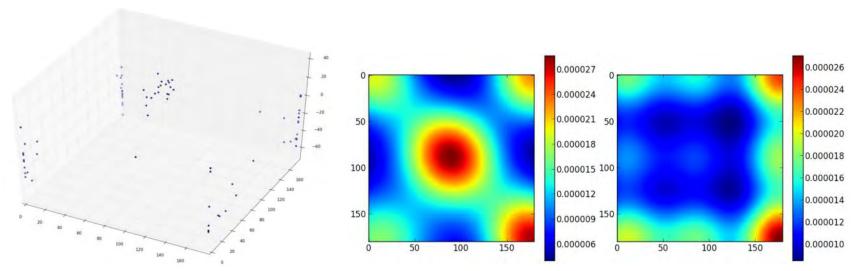
Conditional probabilities between the scales

Properties of new Blobs and Children Blobs

- Number of Blobs
- Positions of Blobs
- Intensities of Blobs

Importance of Multidimensional probability law

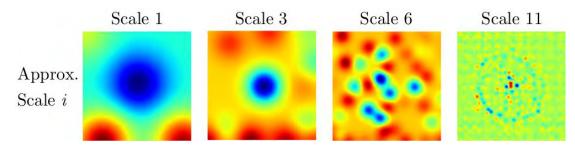






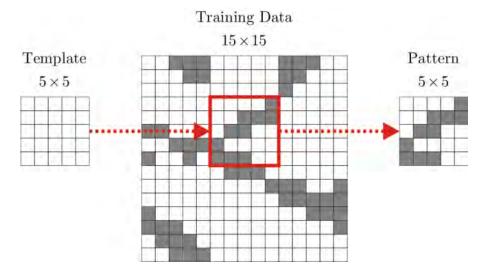
Simulation of Local Structures

Strong increase of number of Blobs for finer structures on higher scales



Assumption: Fine structures have a "local" character

Approach: Analysis and classification of local patterns within moving "template"





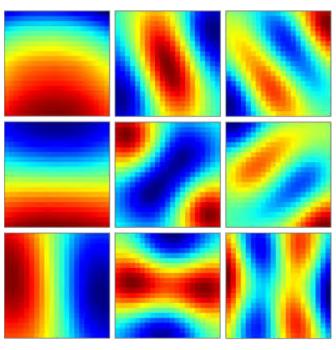
Analysis and Classification of Local Patterns

Relevant local structures have to fit into the template

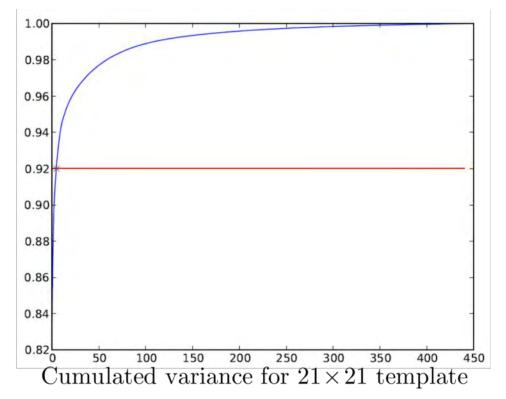
Suitable template size leads to high dimensionality

Treatment in reduced "score space" by means of Principal Component Analysis

14 dimensions instead of 441



Filter 1 ... Filter 9

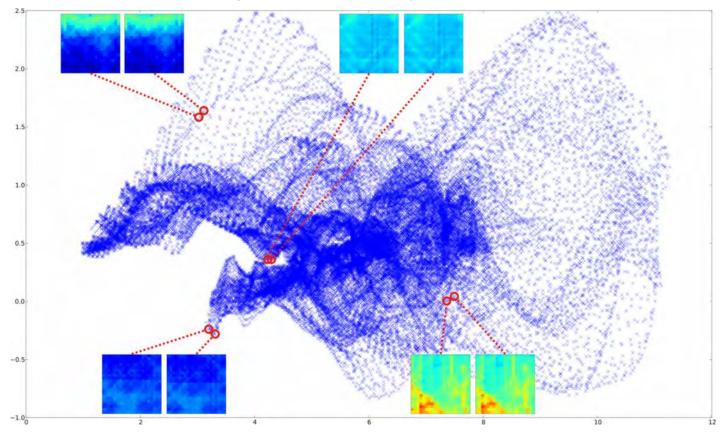




Local Patterns in the Score Space

Scatter plot of principal component 1 and 2 in the score space

- Pattern shows significant amount of structure
- Distance measure in just 2 dimensions yields excellent results



Clustering of Patterns in Score Space

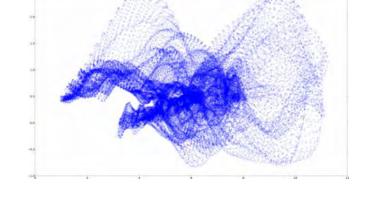
Transformation to score space enables for effective separation of features

- Heterogeneous pattern density in score space: HAC does not yield good results
- Large number of patterns (>10⁶) demands a simple approach

Approach: Binary tree clustering

- Minimum size of cluster
- Minimum "representativeness"

$$R = 1 - \frac{\sum_{i} \sigma_{i}}{n_{\rm class} \sigma_{\rm class}}$$



Class consists of patterns and mean "prototype"





































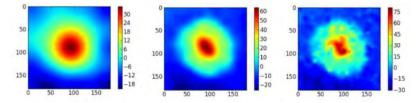




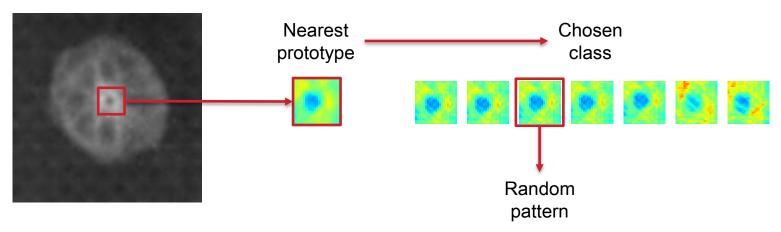
Simulation of Synthetic Realization

Simulation of Synthetic realizations in a two step approach:

1. Simulation of global features and coarse local features by means of multiscalar approach based on conditional probabilities of Blobs between scales

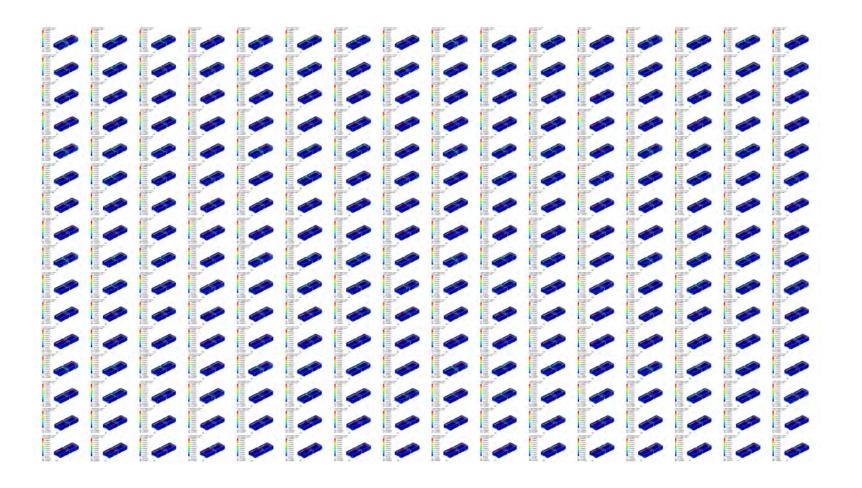


Synthetic realization of Blobs is used as a seed for the completion based on local pattern matching





Simulation Post Processing (Scientific Visualisation)





Limitations of Today's Statistic Visualisation in CAE

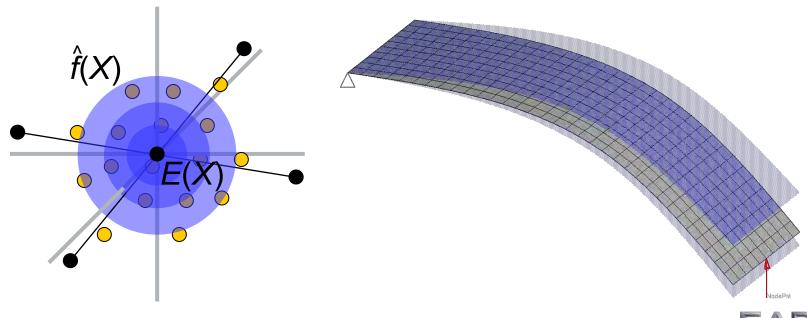
View a single point in space

View a single parameter View at one moment in time $\mu_{\Delta y} \pm \sigma = 2 \pm 5 \text{ mm}$



New Method to Visualise Geometric Scatter

Density estimation of three-dimensional density function from the samples Calculation of the expected value per each node of the mesh Assembly of the expected value geometry by connection of these nodes Derivation of convex hulls with regard to user defined α -levels of maximum probability density



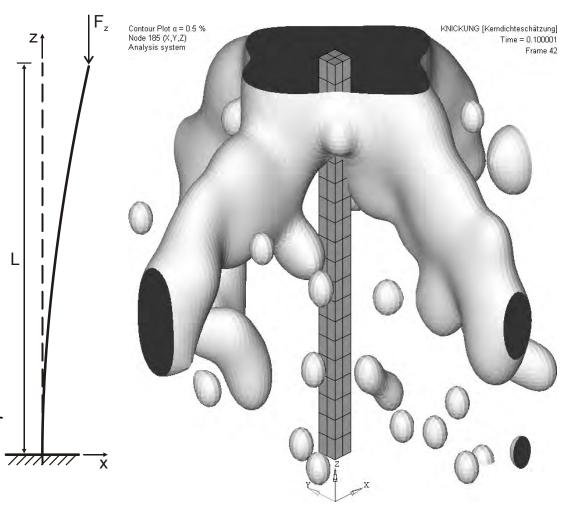
Stochastic Simulation of 1st Euler Buckling Mode

Stochastic parameters:

- Length
- Edge lengths
- Force
- Young's modulus

Isoprobability surface of column tip:

- User defined α-level of max PDF
- Marching cube algorithm for convex hull

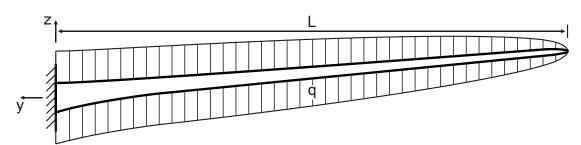


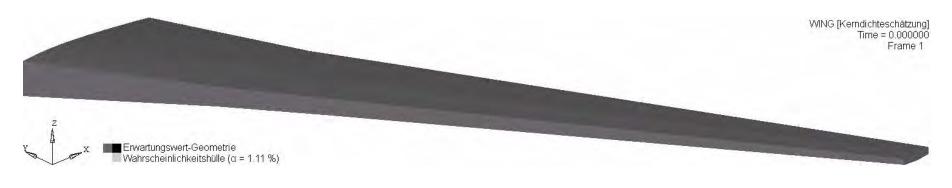


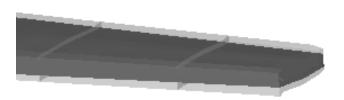
Aircraft Wing Under Varying Load Conditions

Variation of air loads caused by:

- Turbulence
- Altitude
- Flight conditions







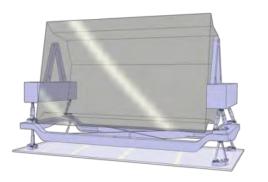


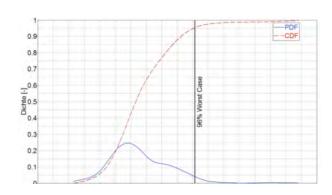
Application Examples

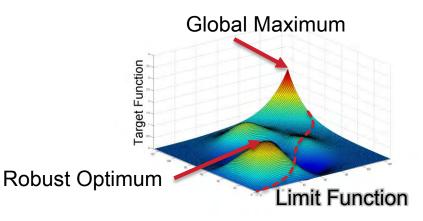
Optimizing models with uncertain system properties or boundary conditions using probabilistic modelling.

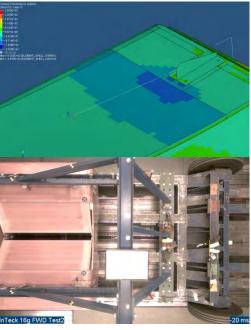
Optimise performance with respect to robustness and reliability

- TRS-3D Shock-Frame with tuned mass-damper
- Energy-Absorber (load limiter) for cabin components









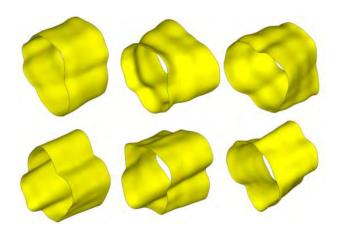


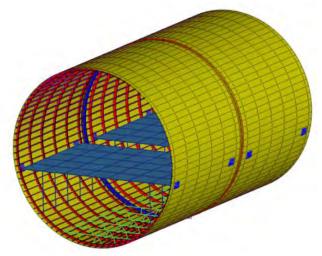
Application Examples

Fuselage assembly process

- Random fields for geometrical uncertainties
- Simulate residual stresses for different riveting sequences





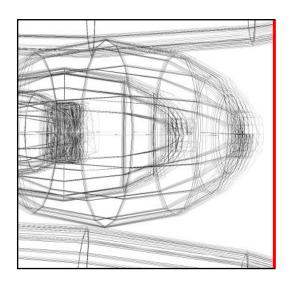




Application Examples

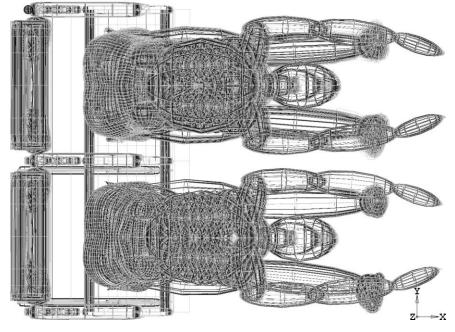
Nonlinear dynamic simulation of seat

- Optimise seat leg stiffness
- Constraint: HIC
- Decrease head displacement



■Expected value geometry
■Isoprobability hull (α = 60,65 %)
■Isoprobability hull (α = 13,53 %)
■Isoprobability hull (α = 1,11 %)
■Isoprobability hull (α = 0,03 %)

FAA-Seat [Kernel density estimate] Time = 0.150000 Frame 15





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Thank you for your attention.



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