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A New Method to Visualize Geometric Scatter from Stochastic Simulation Studies

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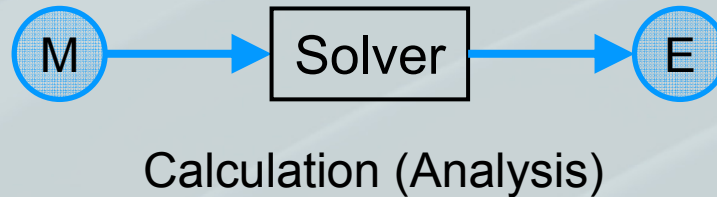
Content

- Introduction to Stochastic Simulation
- Development of Visualisation Method
- Application Examples
- Conclusion

Deterministic Simulation

Problem definition as idealized Model with fixed system properties

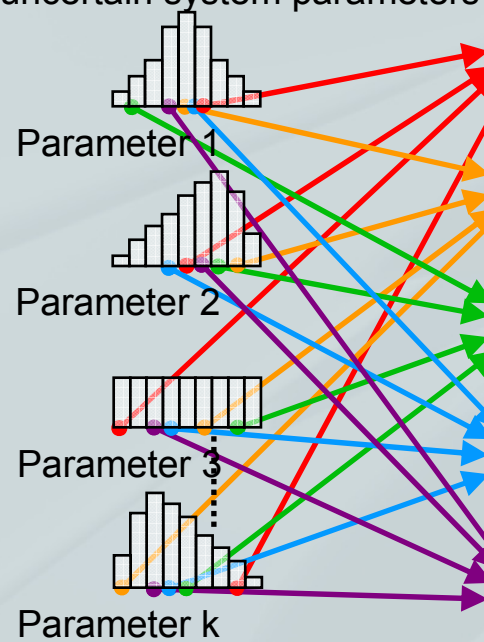
Exact numerical result of the simulation



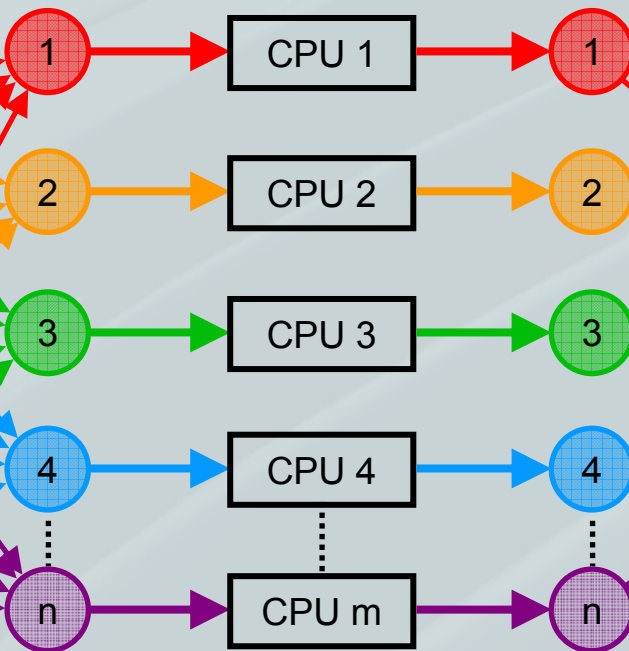
- Fixed link between Model and numerical analysis result
- Disadvantages:
 - Cause-effect relationships of the system may remain concealed
 - Influence of scatter due to production and usage tolerances not quantifiable
 - No information about system robustness

Stochastic Simulation (Monte Carlo Method)

generation of n random samples according to k Input PDFs modelling all uncertain system parameters

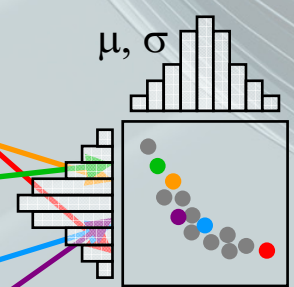


sequential or parallel computation of n simulations on m CPUs



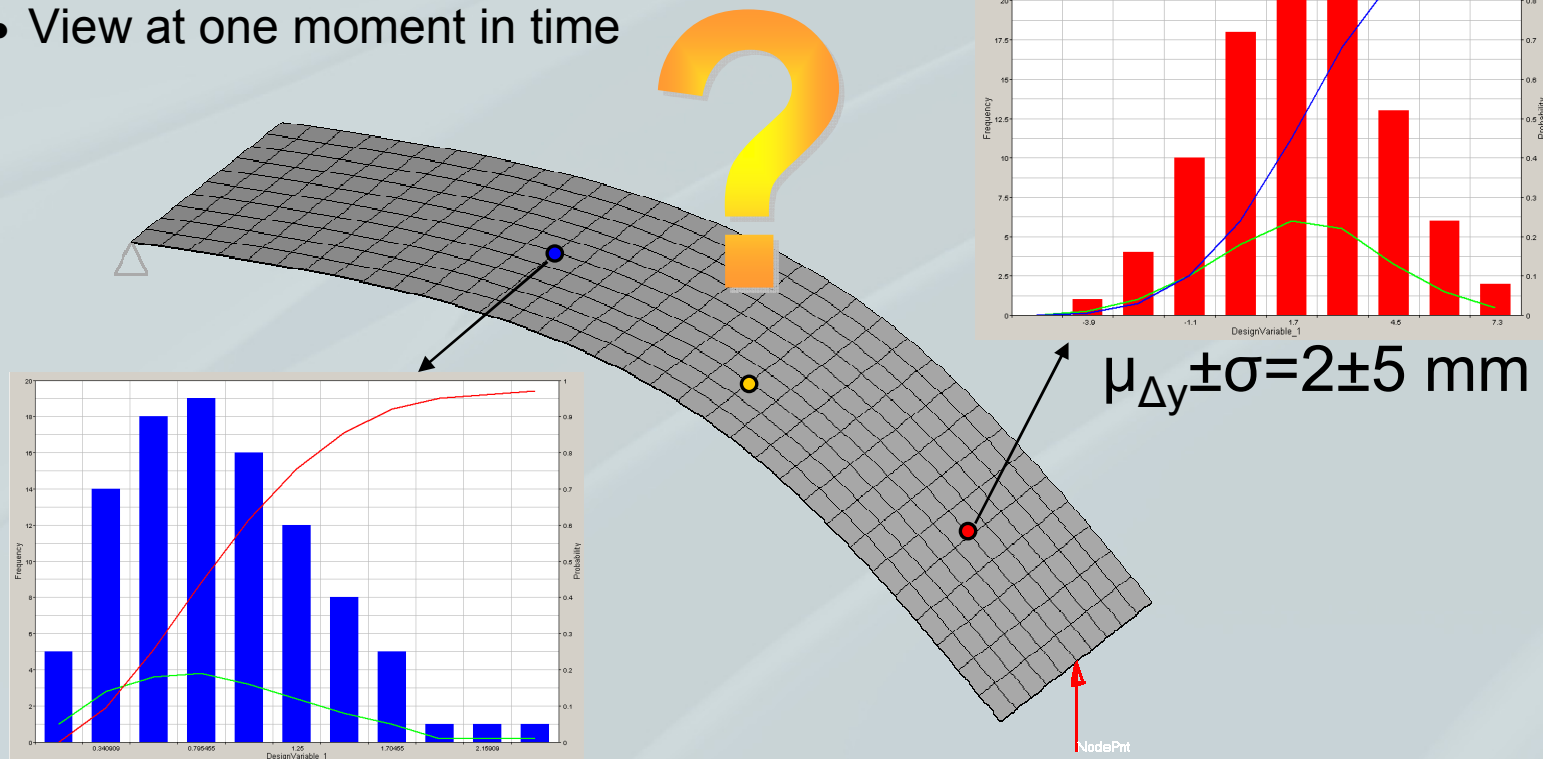
create duplicates of the model for each set of the n random samples

independent to the number of inputs 50 to 150 random samples are required to estimate the global population behaviour



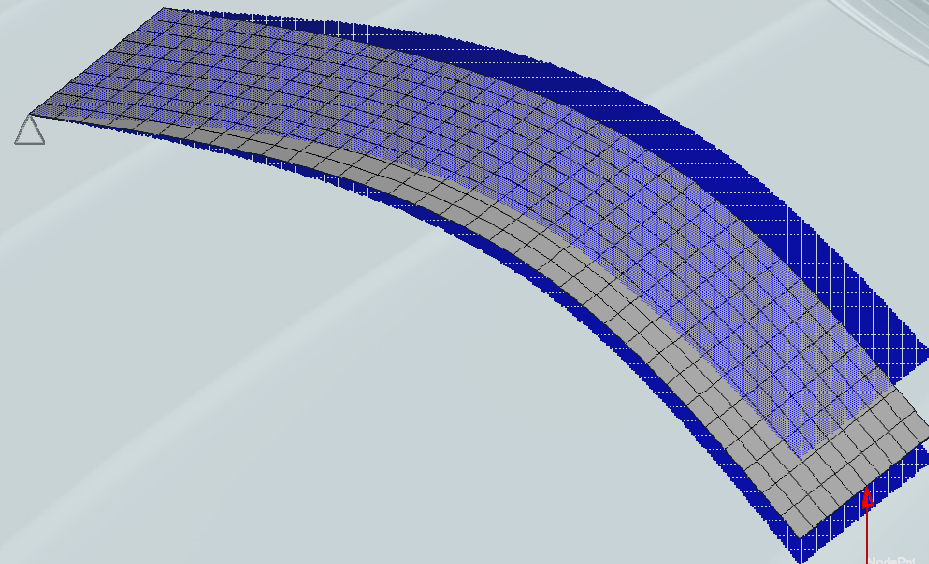
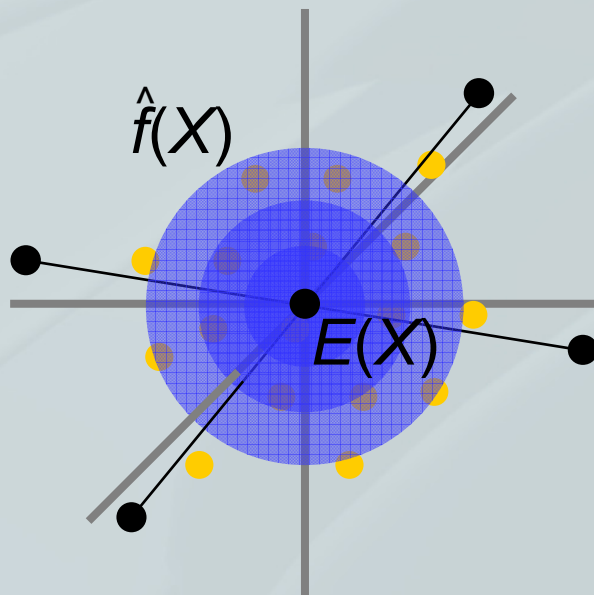
Limitations of today's statistic visualisation in CAE

- View a single point in space
- View a single parameter
- View at one moment in time



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



Non-parametric density estimation

- Kernel density estimator

$$\hat{f}(x) = \frac{1}{n \cdot h} \sum_{i=1}^n K(u)$$

- Epanechnikov-Kernel

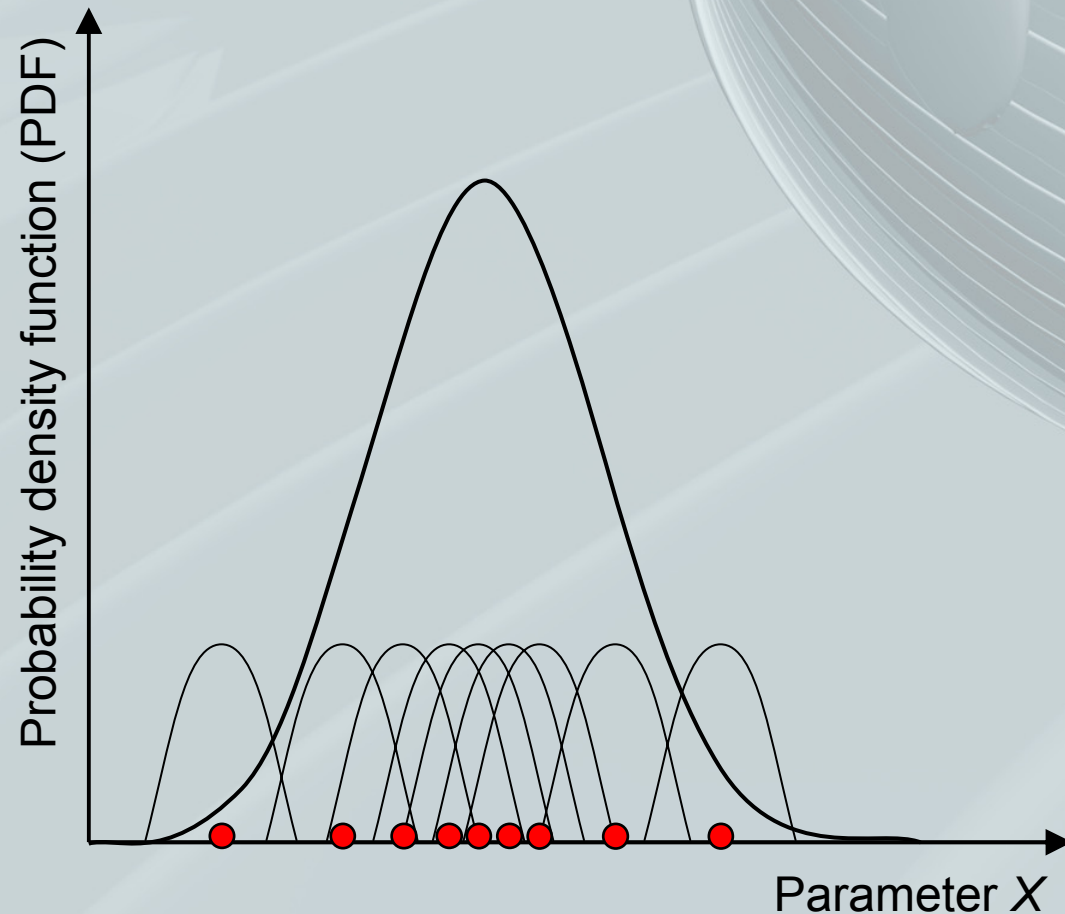
$$K_{\text{Epa}}(u) = \frac{3}{4}(1 - u^2)$$

for $-1 \leq u < 1$,

outside $u = 0$

$$\text{and } u = \frac{x - x_i}{h}$$

- Trade-off between variance and bias depending on bandwidth h



Non-parametric density estimation

- General multivariate kernel density estimator

$$\hat{f}(\mathbf{x}) = \frac{1}{n} \sum_{i=1}^n \left[\prod_{j=1}^d \frac{1}{h_j} K\left(\frac{x_j - x_{ij}}{h_j}\right) \right] \quad \text{mit } x_j \in \mathbf{x}$$

- For three dimensions with Epanechnikov-Kernel

$$\hat{f}(x, y, z) = \frac{3}{4n} \cdot \frac{\sum_{i=1}^n \left[1 - \left(\frac{x - x_i}{h_{\text{Epa}, x}^{\text{opt}}} \right)^2 \right] \cdot \left[1 - \left(\frac{y - y_i}{h_{\text{Epa}, y}^{\text{opt}}} \right)^2 \right] \cdot \left[1 - \left(\frac{z - z_i}{h_{\text{Epa}, z}^{\text{opt}}} \right)^2 \right]}{h_{\text{Epa}, x}^{\text{opt}} \cdot h_{\text{Epa}, y}^{\text{opt}} \cdot h_{\text{Epa}, z}^{\text{opt}}}$$

- Optimum asymptotic bandwidth h

$$\text{with } x_i \in \left(x - h_{\text{Epa}, x}^{\text{opt}}, x + h_{\text{Epa}, x}^{\text{opt}} \right], \quad h_{\text{Epa}, x}^{\text{opt}} = \tilde{s}_x \left(\frac{24\sqrt{\pi}}{5n} \right)^{\frac{1}{7}} \text{ and } y_i, z_i \text{ analog.}$$

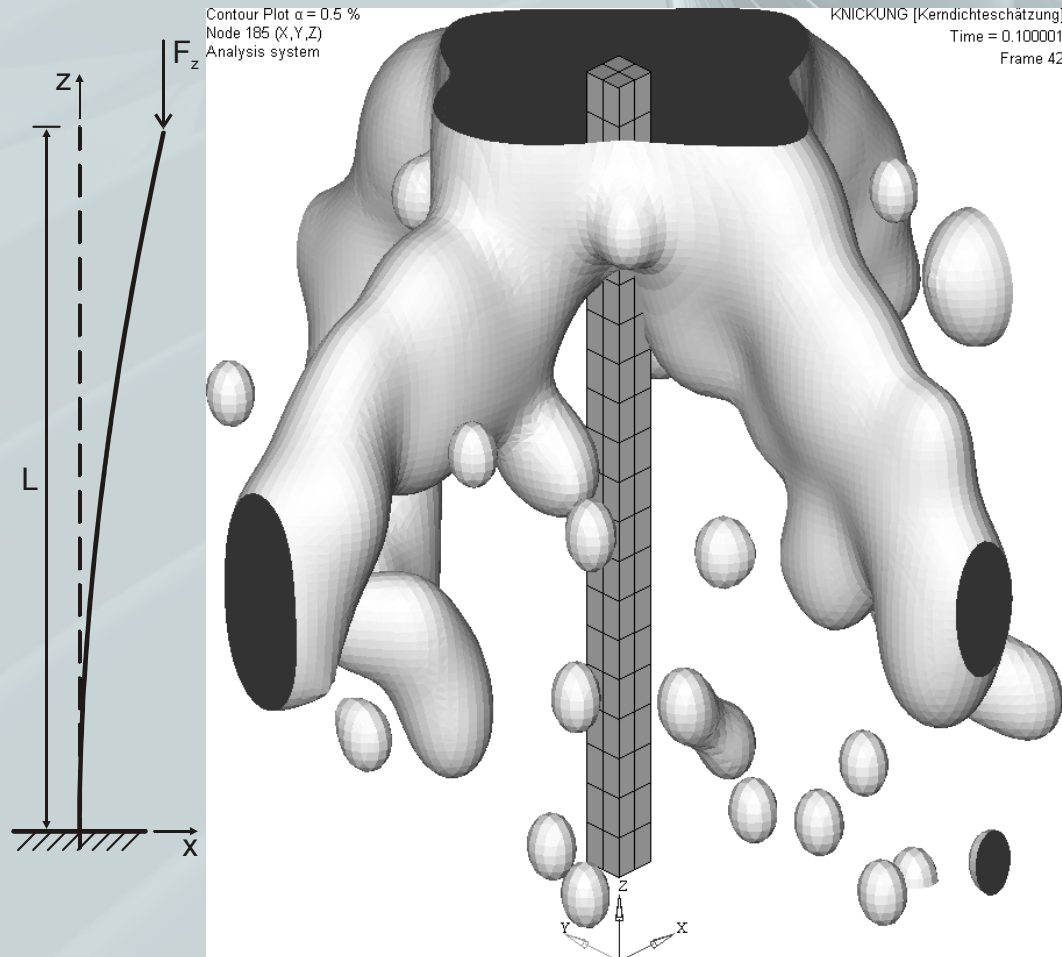
Example: Stochastic Simulation of 1st Euler buckling mode

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

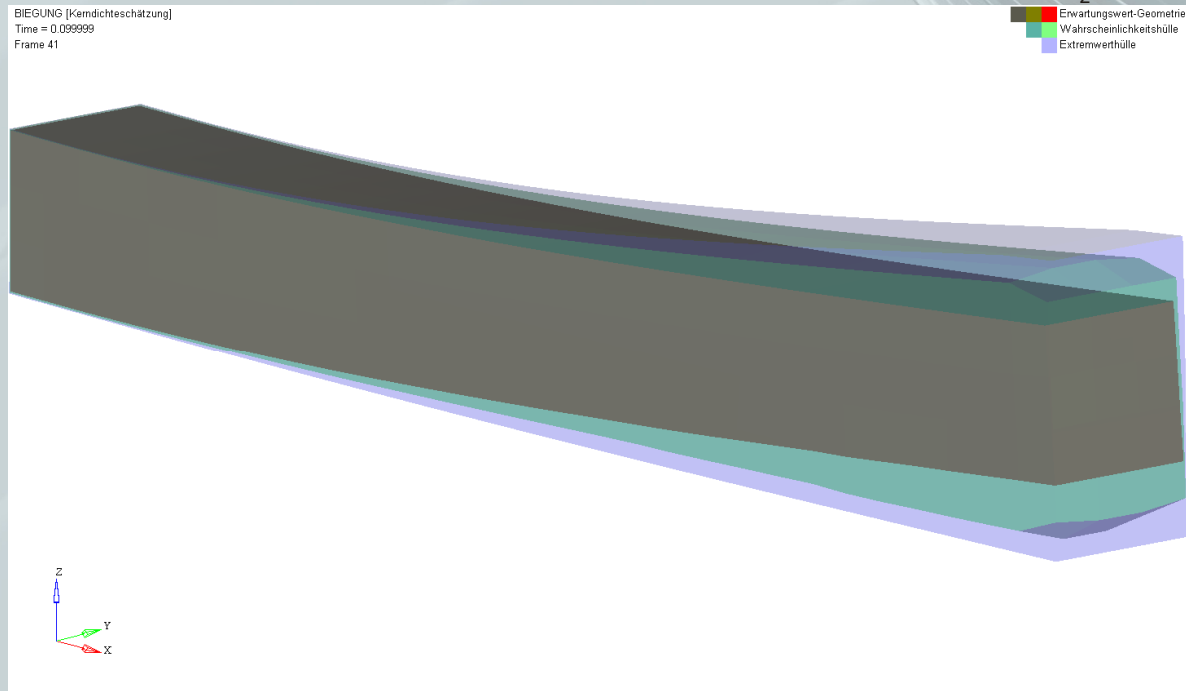
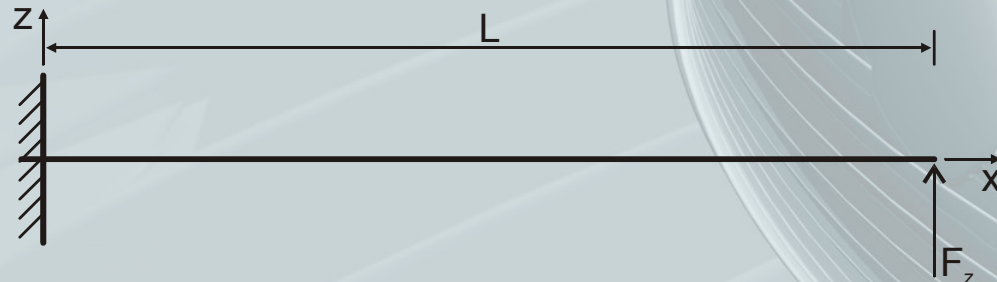


Example: Cantilever with uncertainties

Stochastic parameters:

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

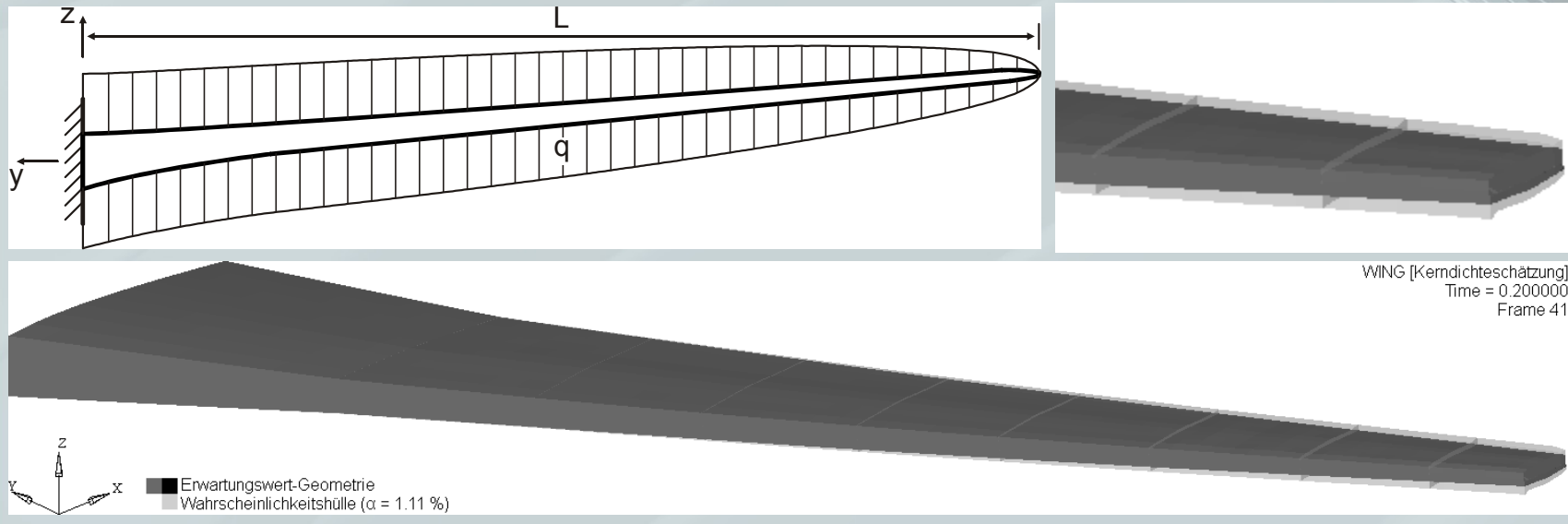
- Expected value geometry with surrounding convex hulls of isoprobability



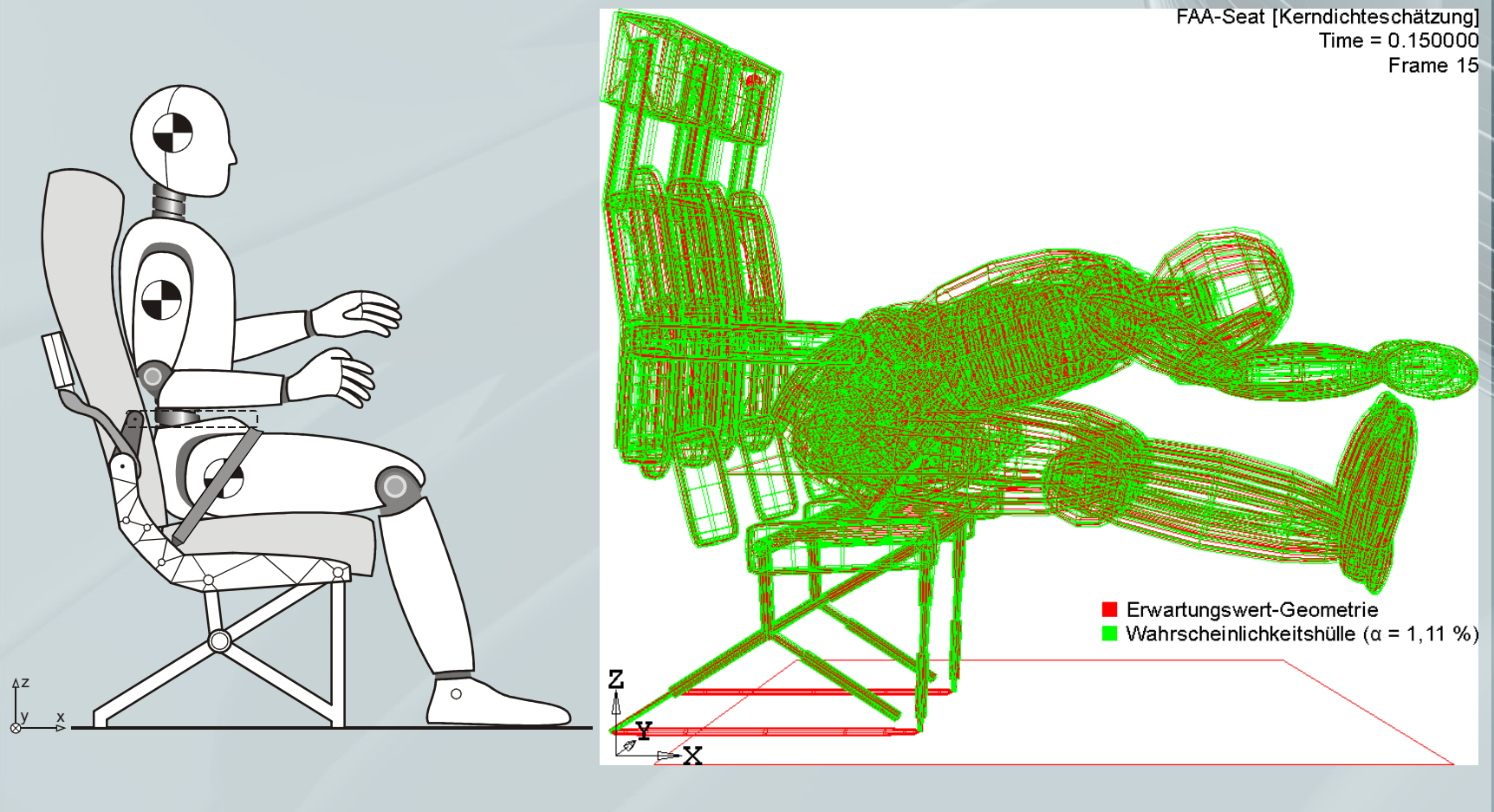
Example: Aircraft wing under varying load conditions

Variation of air loads caused by:

- Turbulence
- Altitude
- Flight conditions



Example: Occupant safety simulation of an aircraft seat



Conclusion

- Assessment of uncertainties important for system reliability and robustness
- Conventional analysis methods limited exploitation of Stochastic Simulation studies

Developed method:

- Relies on non-parametric multivariate density estimation
 - Expected value geometry (most probable system behaviour)
 - Convex hulls of isoprobability considering user defined α -levels
- Provides holistic way to visualise geometric scatter
- Allows for statistic collision detection



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Thank you for your attention.

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