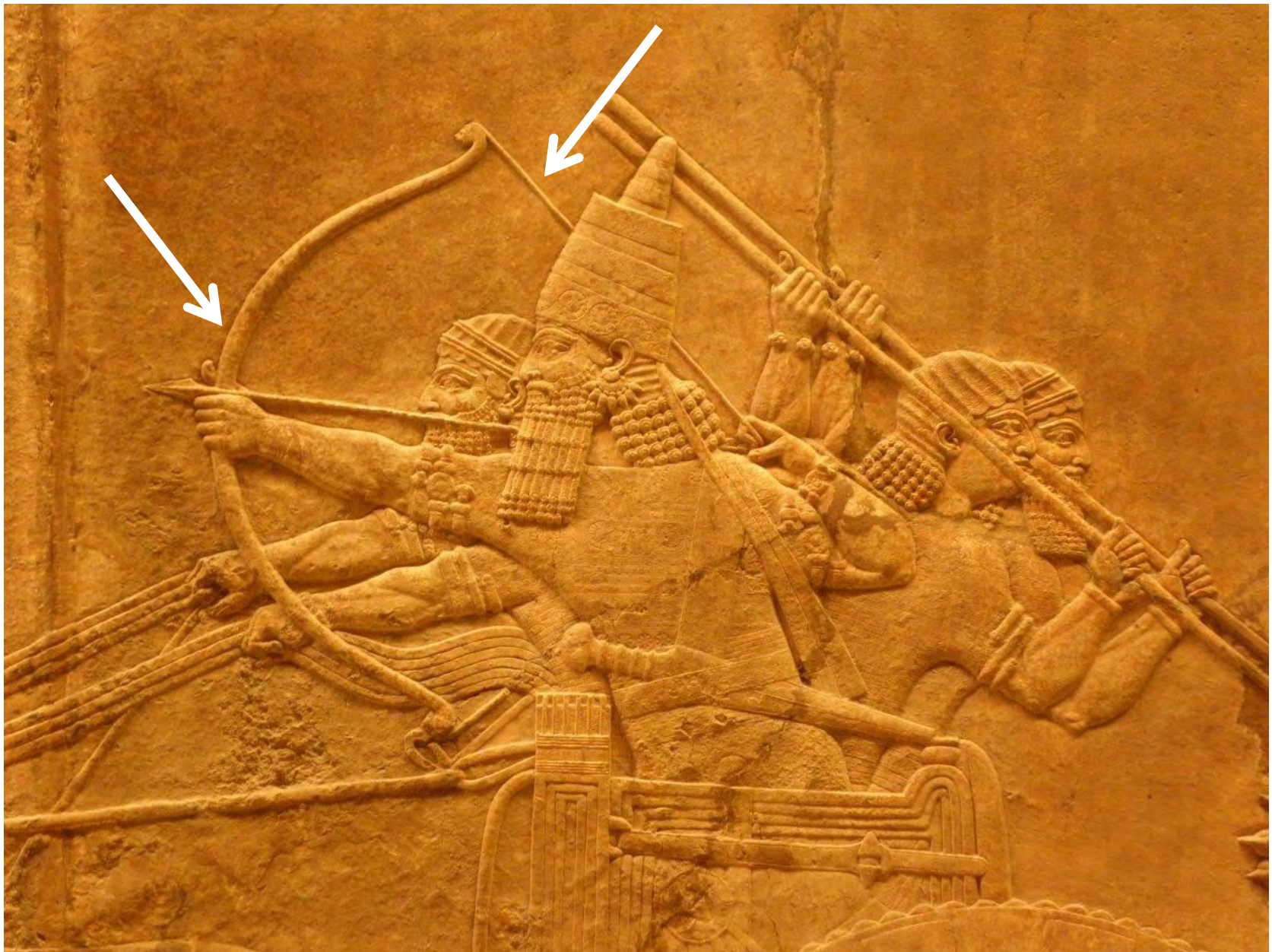


# Quantifying model quality using measured strain fields

Erwin Hack

Empa – Swiss Federal Laboratories for  
Materials Science and Technology





# Bank Stress-Testing, Analysis, and Valuation - London



14–15 October 2013  
London, United Kingdom



London Financial Studies

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## Summary

This is a highly practical and interactive course covering bank stress-testing, analysis, and valuation. It begins with foundation work reviewing the basic bank business model and covering bank accounting and regulation under Basel 3.

Exercises and workshops emphasize the role of stress-testing banks' capital and liquidity, dividing banks that are more susceptible to bank failure and those that are likely to survive. The appropriate selection of gone-concern and going-concern valuation methods is then deployed to value banks' equity and credit.

### Continuing Education Information

As a [provider-level participant](#) in the [Approved-Provider Program](#), London Financial Studies has determined that this program is eligible for continuing education credit. Pending provider confirmation of attendance, CFA participation in this program will be recorded in individual CFA Institute members' CE record.

### Topics

Risk Management



# Content

- **Damage and failure criteria**
- **Strain field measurement in 1D, 2D, and 3D**
- **Data compatibility of data-rich maps**
- **Image decomposition methods**
- **Quantification of model quality**
- **Conclusions**



# Stress, strain and damage

- **Many fracture/yield/plasticity criteria are based on stress (or strain) values**
  - von Mises, Tresca, Puck, **Logan-Hosford**

$$F|\sigma_2 - \sigma_3|^n + G|\sigma_3 - \sigma_1|^n + H|\sigma_1 - \sigma_2|^n = \sigma_y^n$$

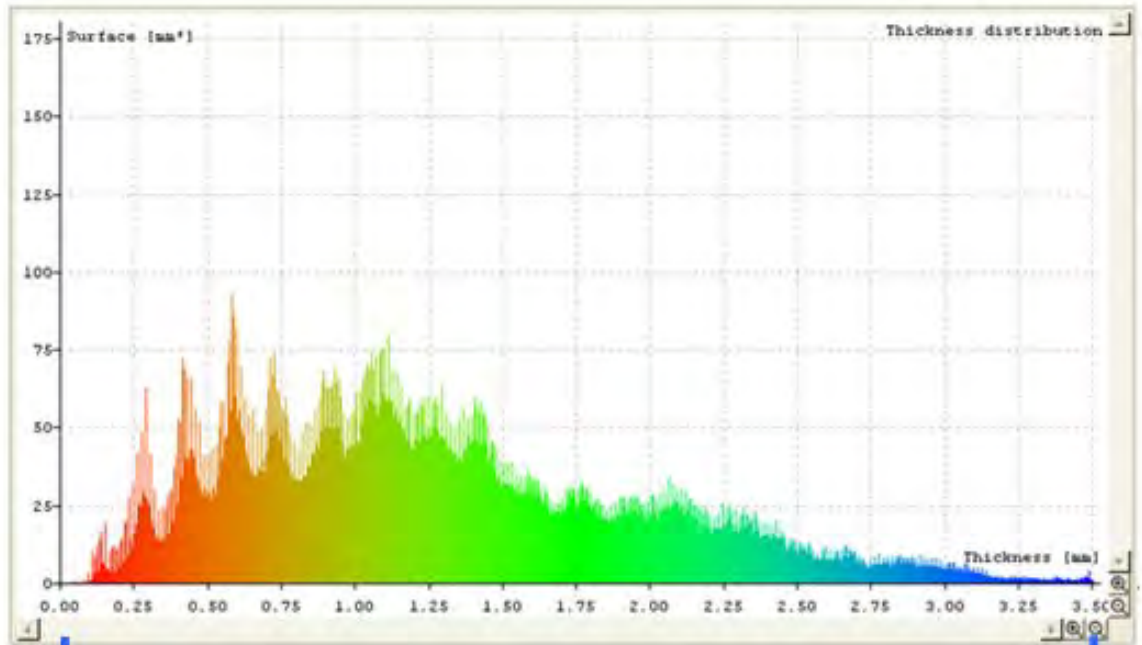
- **Local stress values are difficult to obtain for a component under service load**
- **Use defined stress cases in model systems first**





# Volume data

- distribution of voids from x-ray CT



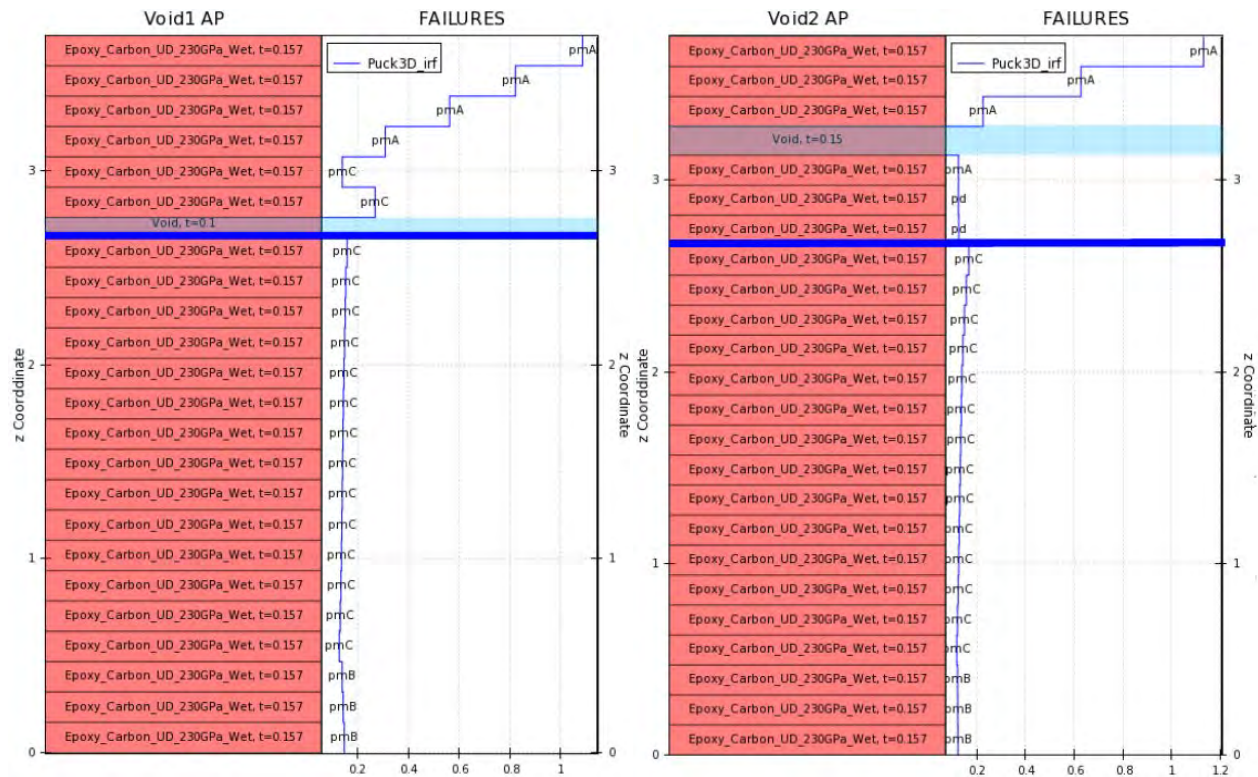
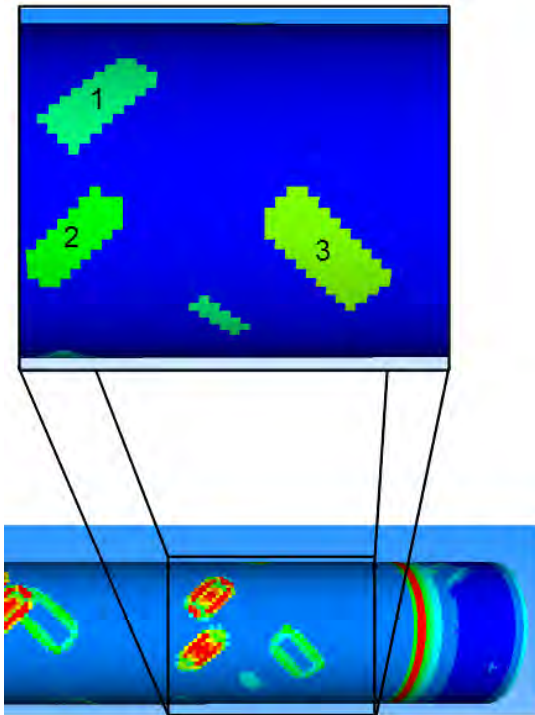
External layer

Internal layer

E. Hack, M. Feligiotti, R.K. Fruehmann, and J.M. Dulieu-Barton, *Failure and damage in CFRP torsion tubes*, Photomechanics, Montpellier, 27-29 May 2013



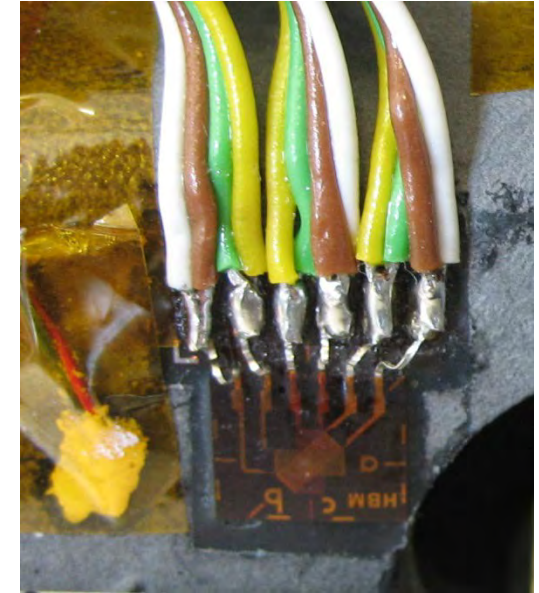
# 3D-model for damage prediction



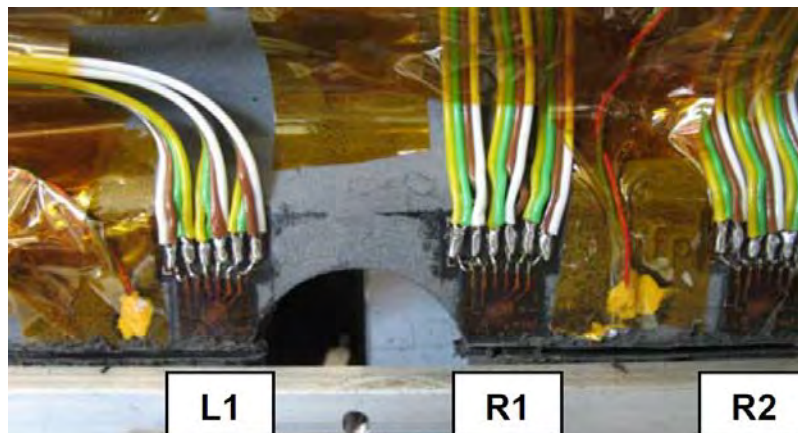


# Case study: FRP injection moulded component

Specimen with direction of cracks induced by residual stress



Mount for saw notch method

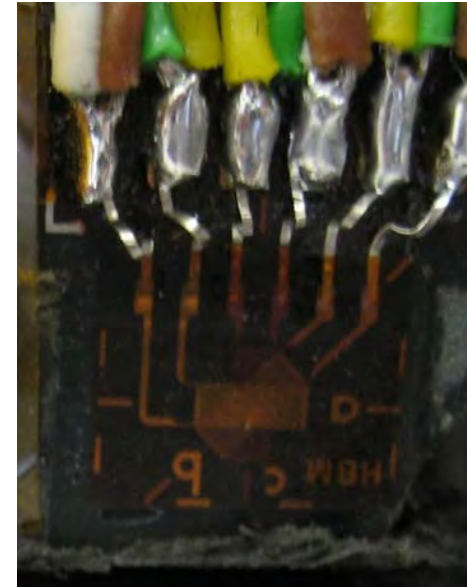
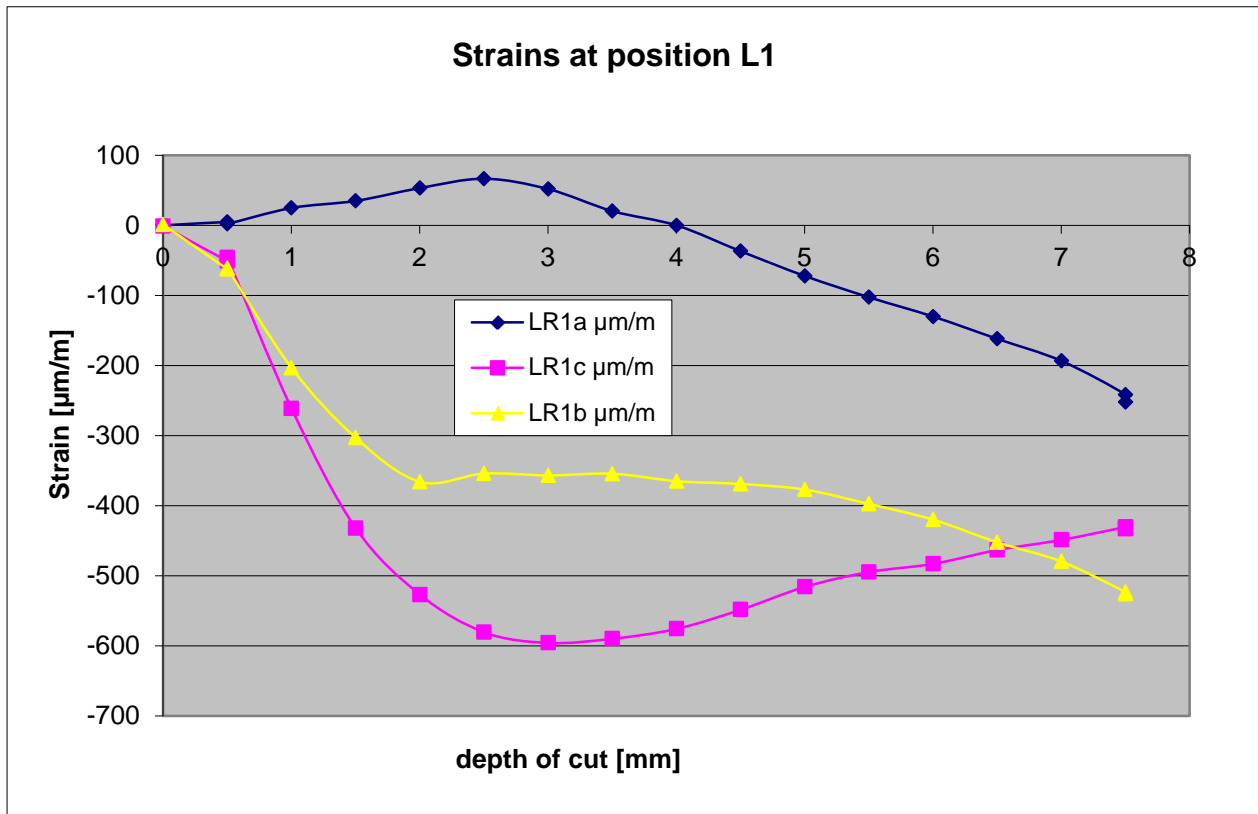


Local measurement of surface strains with RSG-Rosette



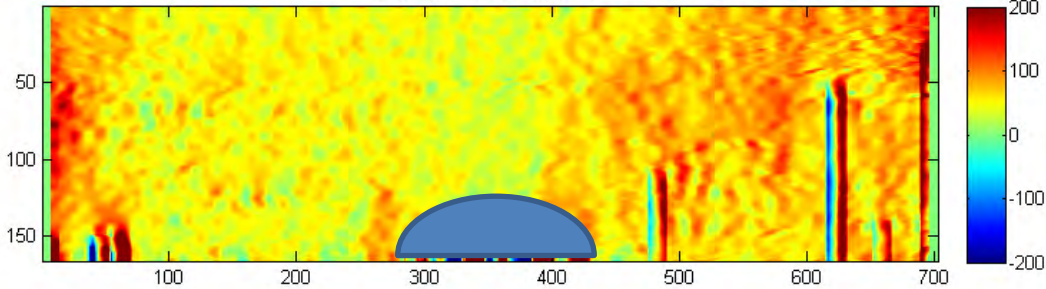
# RSG: local strain data

- Measured at increasing notch depth
  - after cooling down (thermocouple)

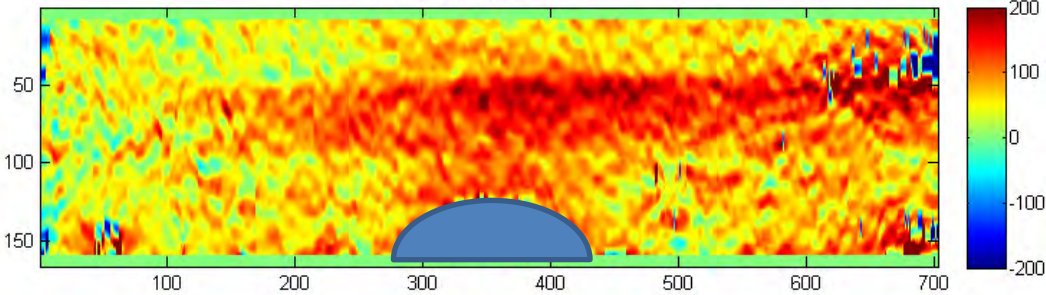


# DSPI: global strain data

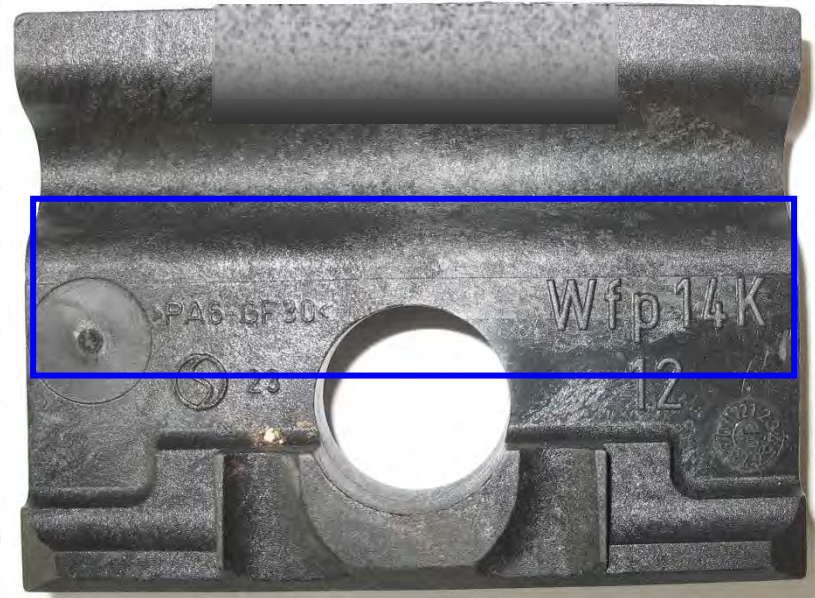
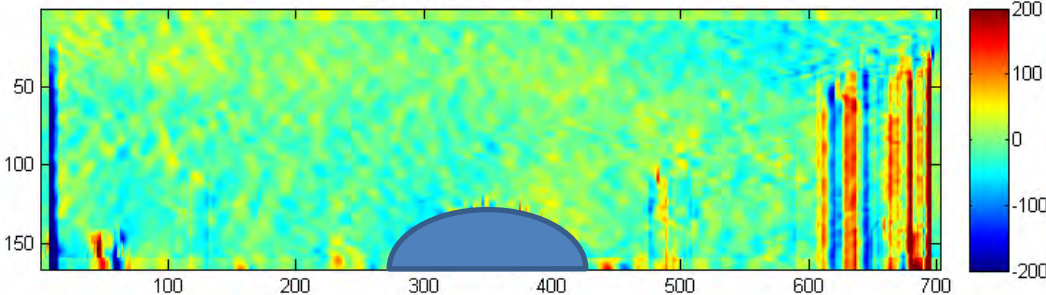
x-strain [ $\mu\text{m}/\text{m}$ ], Sample 3a



y-strain [ $\mu\text{m}/\text{m}$ ], Sample 3a



xy-strain [ $\mu\text{m}/\text{m}$ ], Sample 3a

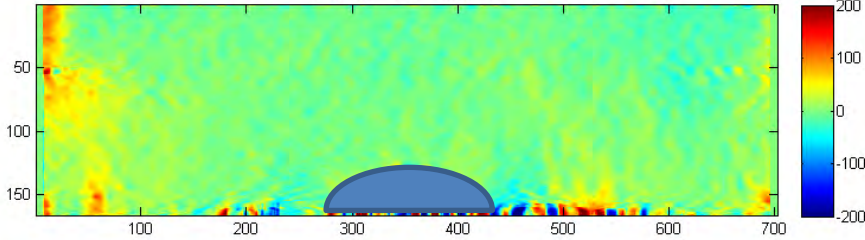


Strain values for saw notch  
0.0 – 0.5 mm

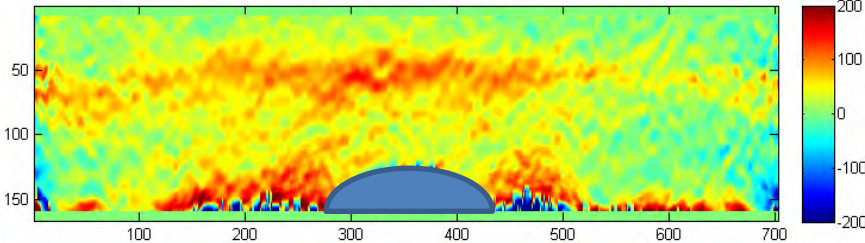


# DSPI: strain reversal

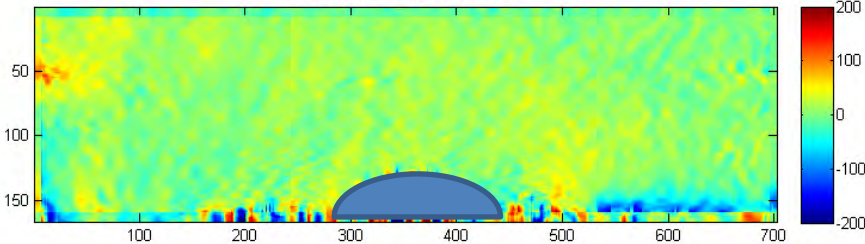
x-strain [ $\mu\text{m}/\text{m}$ ], Sample 3a



y-strain [ $\mu\text{m}/\text{m}$ ], Sample 3a

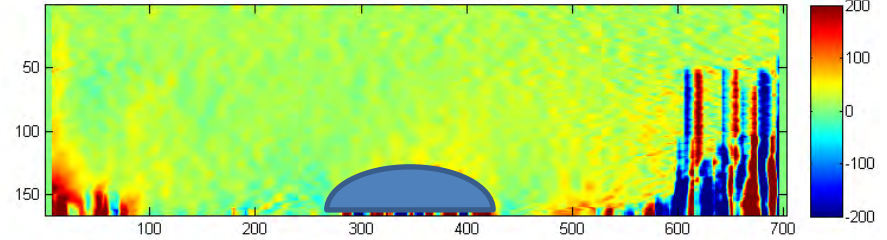


xy-strain [ $\mu\text{m}/\text{m}$ ], Sample 3a

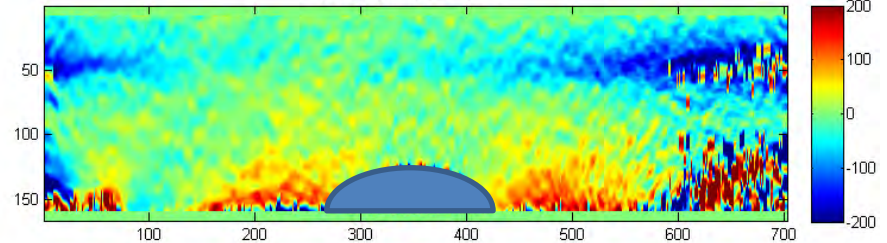


Strain values for saw notch  
0.5 – 1.0 mm

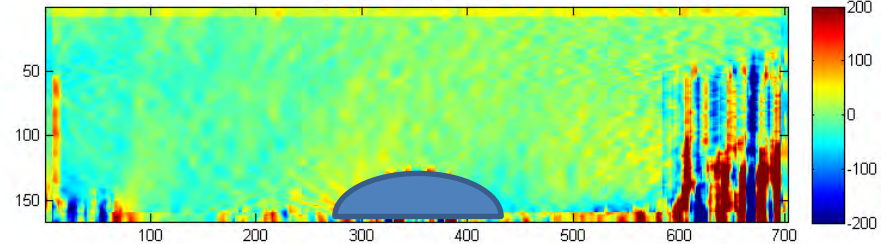
x-strain [ $\mu\text{m}/\text{m}$ ], Sample 3a



y-strain [ $\mu\text{m}/\text{m}$ ], Sample 3a



xy-strain [ $\mu\text{m}/\text{m}$ ], Sample 3a



Strain values for saw notch  
1.0 – 1.5 mm



# Surface strain

- Since x-ray CT cannot be used outside the lab volume data and volume strain are not accessible
- Yet measurement of surface strains is feasible, *in situ* and under defined load (stress)
- Some 3D information can be obtained destructively
- Commonly therefore, surface strain values are compared with numerical model data



# Modeling quality

- Numerical 3D-models calculate strain/stress fields and explain failure load and mode
- **What are model quality issues?**
  - Appropriate physics
  - Code verification
  - Appropriate meshing
  - Appropriate boundary conditions
  - Convergence of solution
  - **Robustness of solution**
  - **Correctness of solution**



# P R E F A C E

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THANKS to the labors of Kirchhoff, Kelvin, Huxley, and others, there is now a widespread opinion that any physical phenomenon is “explained” only when some one has devised a dynamical model which will duplicate the phenomenon. The completeness of the explanation is to be measured by the completeness with which the model will duplicate the phenomenon.

Henry Crew (Editor), *The wave theory of light: memoirs by Huygens, Young and Fresnel*, American Book Company, New York 1900.

- This statement refers to a physical theory, but...
- could a similar statement hold true for numerical models?



# Faithful representation

- **«Devise a dynamical model»**
  - set-up a numerical model
- **«Duplicate the phenomenon»**
  - explain experimental outcome
- **«Completeness of explanation»**
  - quantify deviations

THANKS to the labors of Kirchhoff, Kelvin, Huxley, and others, there is now a widespread opinion that any physical phenomenon is “explained” only when some one has devised a dynamical model which will duplicate the phenomenon. The completeness of the explanation is to be measured by the completeness with which the model will duplicate the phenomenon.



# Faithfulness and validation

- **VIM definition (JCGM 200:2012)**
- **Validation** is defined as the provision of objective evidence that a given item fulfils specified requirements adequate for an intended use.
- Intended use: **Fitness for purpose**



# «Duplicate the phenomenon»: Comparison of model and measurement data

- **General concept** in analogy to **metrological compatibility of measurement results**
- **VIM 2012:**  
“property of a set of **measurement results** for a specified **measurand**, such that the absolute value of the **difference of any pair of measured quantity values** from two different measurement results is smaller than some **chosen multiple of the standard measurement uncertainty** of that difference”

$$|\varepsilon_{1n} - \varepsilon_{2m}| \leq \kappa \times u(d) \quad \forall n, m$$

**Single measurement value criterion!**



# Point-wise comparison of data fields

$$d(\mathbf{x}_M) = \varepsilon_{opt}(\mathbf{x}_M) - \varepsilon_{FEM}(\mathbf{x}_M)$$

- Data sets are expressed as  $N$ -dimensional vectors
- An overall quantitative quality criterion is a must
  - e.g. when different FEA results based on different models or different parameter values are available.
- The following slide shows an indicative list of possible quality criteria (“cost functions”, “distance measures”).
  - All summations are weighted summations, but we suppress the weighting factors for clarity.



# Quality criteria for a set of points

• max. allowable deviation:  $Max = \max(|d(\mathbf{x}_M)| | M = 1 \dots N)$

• rms criterion:  $rms = \frac{1}{N} \sqrt{\sum_{M=1}^N d^2(\mathbf{x}_M)}$

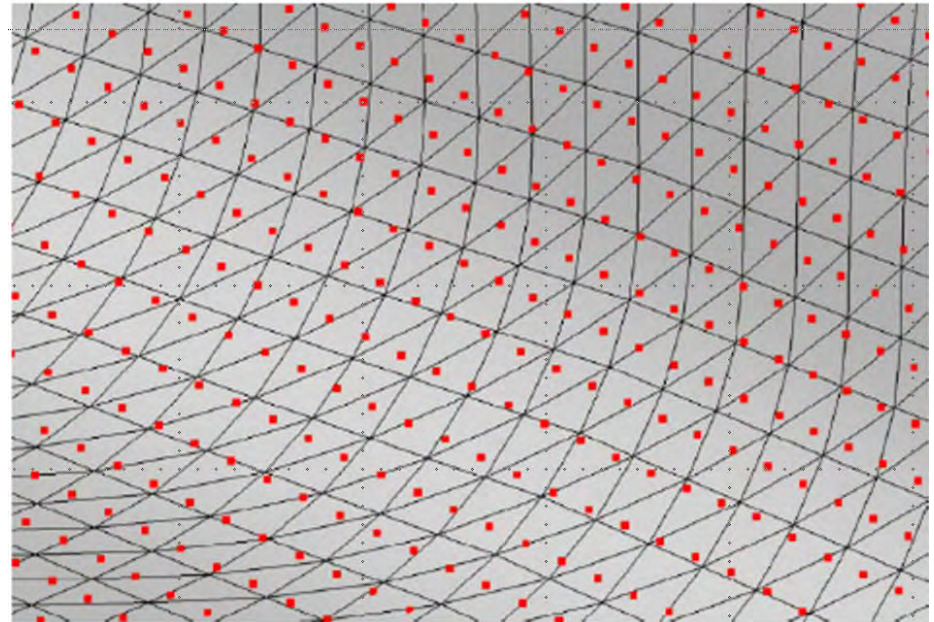
• Modal Assurance Criterion:  $MAC = \frac{\left| \sum_{M=1}^N \varepsilon_{opt}(\mathbf{x}_M) \cdot \varepsilon_{FEM}(\mathbf{x}_M) \right|^2}{\sum_{M=1}^N |\varepsilon_{opt}(\mathbf{x}_M)|^2 \sum_{M=1}^N |\varepsilon_{FEM}(\mathbf{x}_M)|^2}$

• Normalised cross correlation:  $r = \frac{N \sum_{M=1}^N \varepsilon_{opt}(\mathbf{x}_M) \cdot \varepsilon_{FEM}(\mathbf{x}_M) - \sum_{M=1}^N \varepsilon_{opt}(\mathbf{x}_M) \sum_{M=1}^N \varepsilon_{FEM}(\mathbf{x}_M)}{\sqrt{N \sum_{M=1}^N |\varepsilon_{opt}(\mathbf{x}_M)|^2 - \left( \sum_{M=1}^N \varepsilon_{opt}(\mathbf{x}_M) \right)^2} \sqrt{N \sum_{M=1}^N |\varepsilon_{FEM}(\mathbf{x}_M)|^2 - \left( \sum_{M=1}^N \varepsilon_{FEM}(\mathbf{x}_M) \right)^2}}$



# Point-wise comparison

- Comparison can be performed point by point and the criterion be applied
- Involves step of point matching and data interpolation



Measurement points (red)  
vs. FEA grid (lines)

# Unequal number of points?

- **Point-to-point comparison requests**  $N_{opt} = N_{FEM}$ 
  - Missing data points generated by intra-/extrapolation
- **Is there a different way of comparing unequal sets of data?**
- **Idea: Parameterize field of data points and compare the reduced data**

$$\begin{aligned}\varepsilon_{opt}(x, y) &= \sum_{nm} a_{nm} f_n(x) f_m(y) \\ \varepsilon_{FEM}(x, y) &= \sum_{nm} b_{nm} f_n(x) f_m(y)\end{aligned}$$
$$\varepsilon_{opt}(x, y) - \varepsilon_{FEM}(x, y) = \sum_{nm} (a_{nm} - b_{nm}) f_n(x) f_m(y)$$



# Orthogonal basis systems

- For circular domain : **Zernike** polynomials
- For rectangular domain:
  - **Fourier** components
  - **Orthonormal polynomials**
  - Discrete version
  - e.g. discrete Chebyshev polynomials

$$\int f_m(x)f_n(x)W(x)dx = \delta_{mn}\rho_n$$

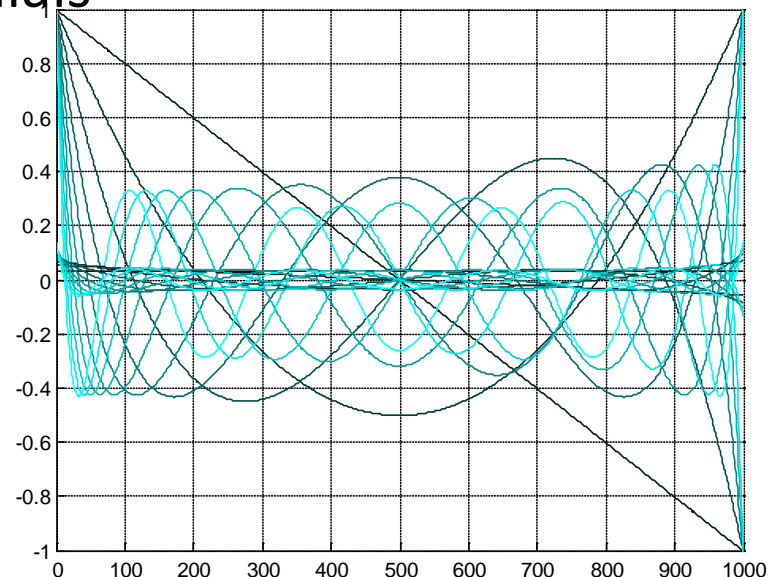
$$\sum_{p=0}^{N-1} f_m(p)f_n(p)W(p) = \delta_{mn}\rho_n$$

$$T_0(p) = 1 \quad p \in [0, N-1]$$

$$T_1(p) = 1 - \frac{2p}{N-1}$$

$$T_{n+2}(p) = \frac{1}{(n+1)(N-1-n)} [(2n+1)(N-1-2p)T_{n+1}(p) - n(n+N)T_n(p)]$$

Discrete Chebyshev polynomials



# Normalisation issue

- The coefficients should not depend (explicitly) on number of points.

$$\varepsilon(p) = \sum_{n=0}^P a_n f_n(p)$$

$$a_n = \sum_{p=0}^{N-1} \varepsilon(p) f_n(p)$$

- If  $\varepsilon(p)$  is a constant

$$a_0 = \sum_{p=0}^{N-1} \varepsilon(p) f_0(p) = \frac{1}{N} \sum_{p=0}^{N-1} \varepsilon(p)$$

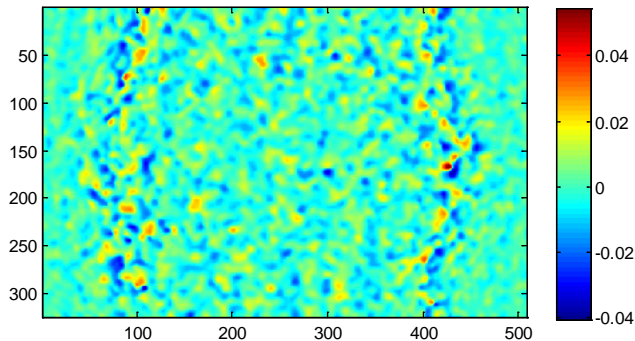
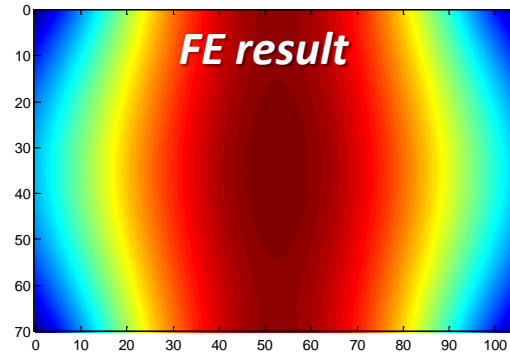
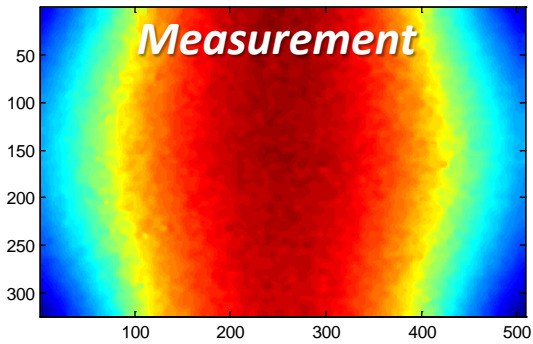
$$\sum_{p=0}^{N-1} f_m(p) f_n(p) = \delta_{mn} \rho_n$$

thus  $f_0(p) = \frac{1}{N}$  and  $\rho_0 = \frac{1}{N}$

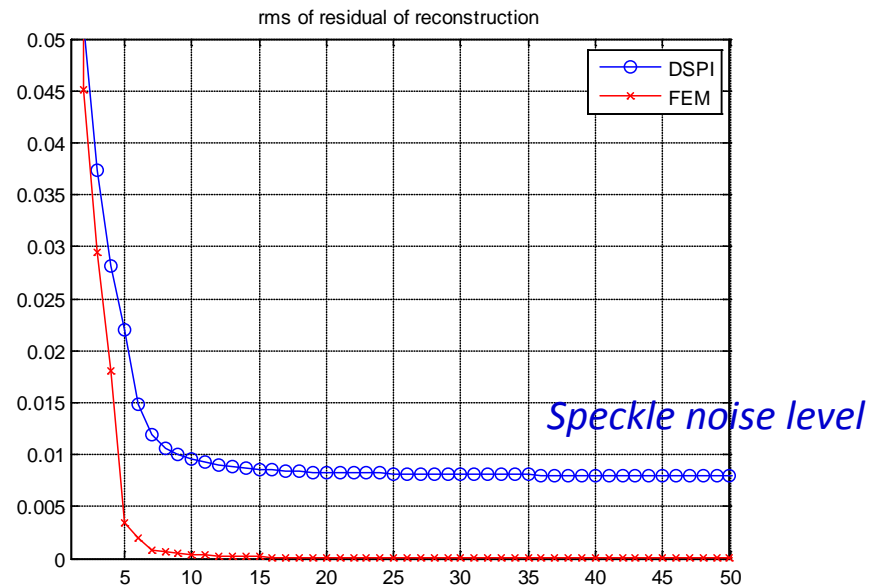




# Image decomposition



*Residuals for a 20 term reconstruction of DSPI map*



*Mean square residual after reconstruction*



# Methodology to compare experimental and simulation data

$d(\mathbf{x}_M) = \varepsilon_{opt}(\mathbf{x}_M) - \varepsilon_{FEM}(\mathbf{x}_M)$  is replaced by  $d_{nm} = a_{nm} - b_{nm}$

$M = O(10^6)$  points

$nm = O(10^3)$  coefficients

- The basis functions for *FEM* and *opt* are calculated on the respective domains
- The coefficients  $a_{nm}$  and  $b_{nm}$  have the same dimension as the measurand, viz. strain
- To display the differences it is best to sort the coefficients according to their value



# Compatibility of data

- based on the *standard measurement uncertainty*

$$d_{nm} = a_{nm} - b_{nm}$$

$$u^2(d_{nm}) = u^2(a_{nm}) + u^2(b_{nm})$$

- It can be shown that for orthonormal decomposition the uncertainties of the coefficients are equal for all  $nm$ .

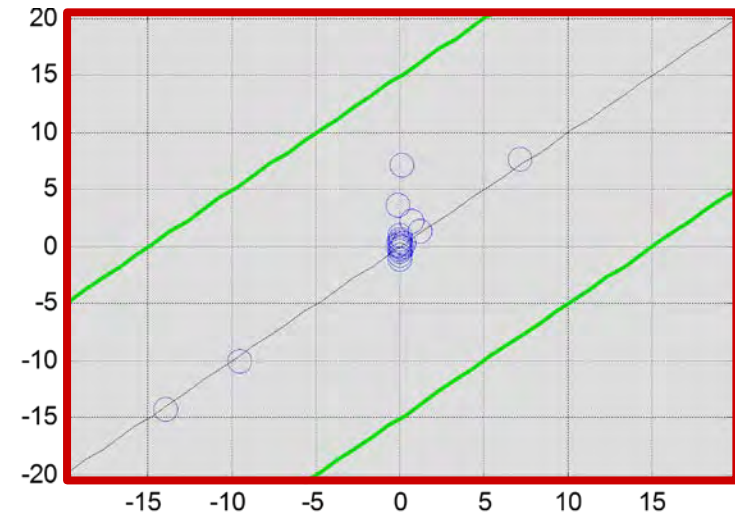
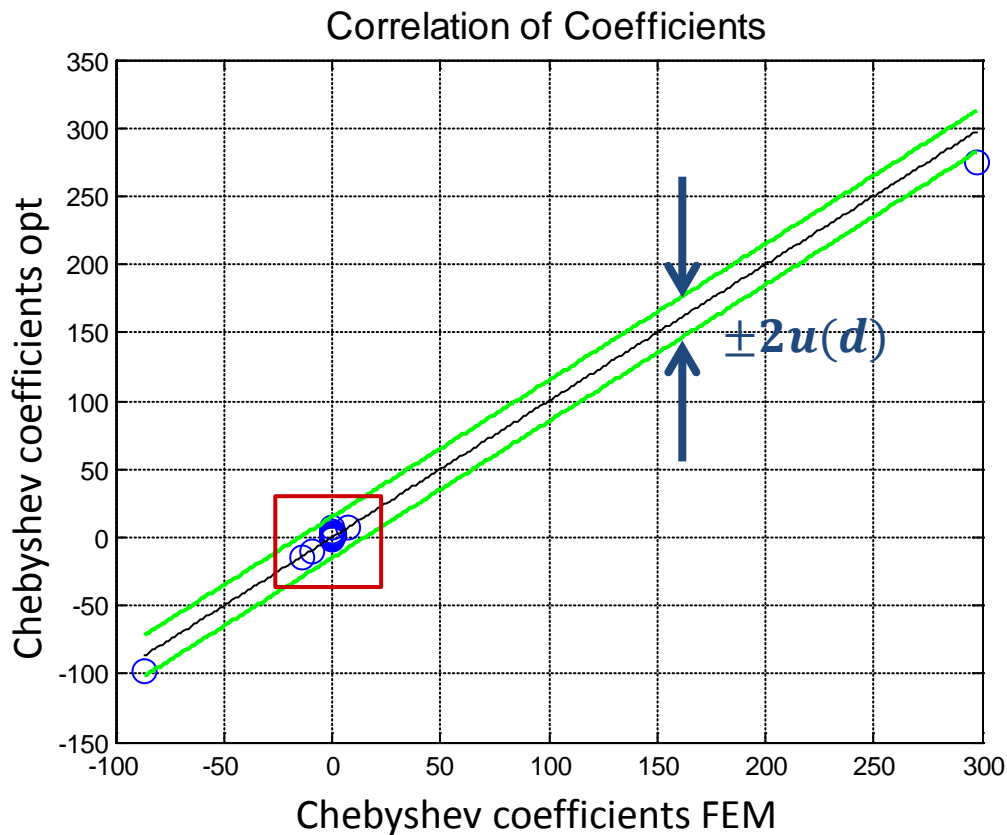
$$u^2(d) = u^2(a) + u^2(b)$$

$$u^2(d) = u_{opt}^2 + u_{FEM}^2$$

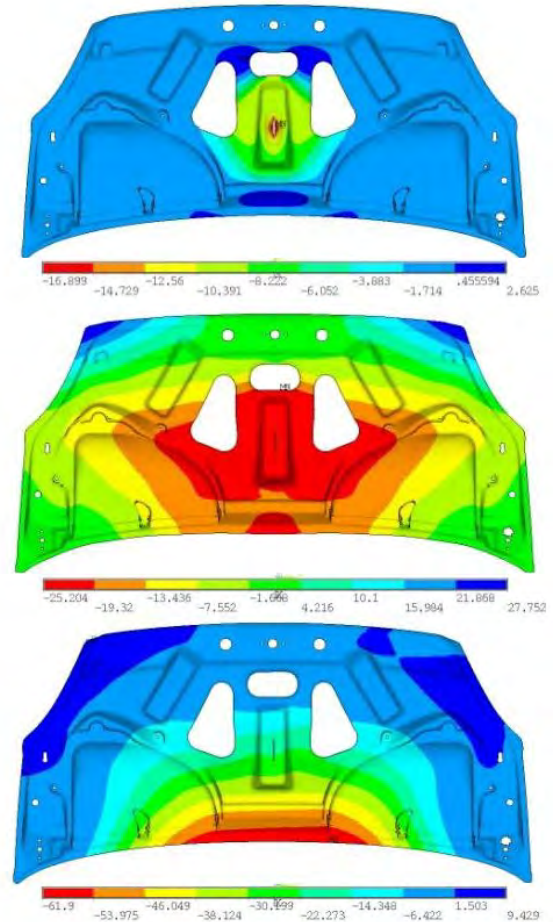


# Comparison of coefficients

- Orthonormal Chebyshev decomposition



# Prospect: non-rectangular domains



W. Wang and J.E. Mottershead, *Adaptive moment descriptors for full-field strain and displacement measurement*, J Strain Analysis 48 (2012) 16-35



# Summary

- **Motivated the use of full-field (surface) strain data based on failure criteria**
- **Viewed model validation in analogy to metrological compatibility of measurement results**
- **Suggested a method to compare full-field data sets by data reduction**
- **Suggested a criterion to quantify model quality for a validation test**



# Acknowledgement

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  - University of Liverpool (UK)
  - University of Patras (Greece)
  - Michigan State University (USA)
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