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Computationally efficient stress reconstruction for nonlinear VFM

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Who we are...



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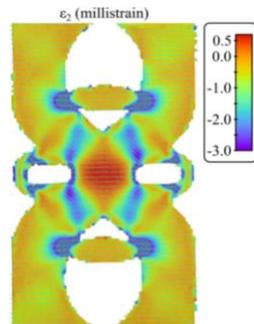
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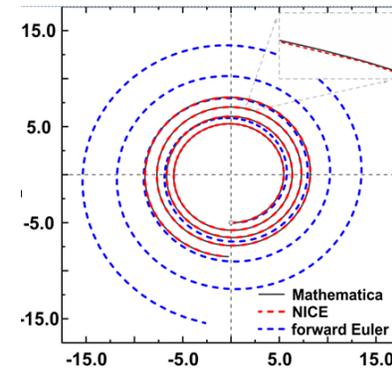


Material modelling & characterization

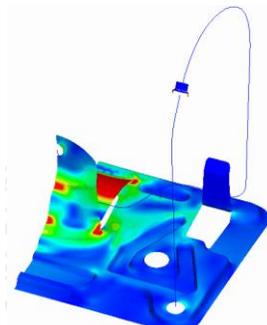
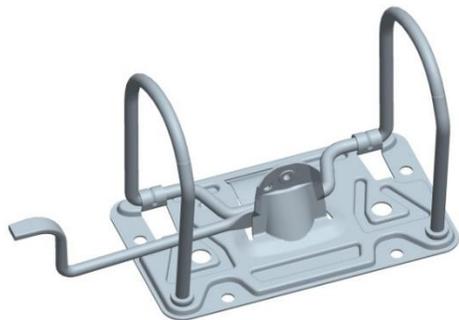


Numerical methods & implementations of material models

- metals
- polymers
- composites

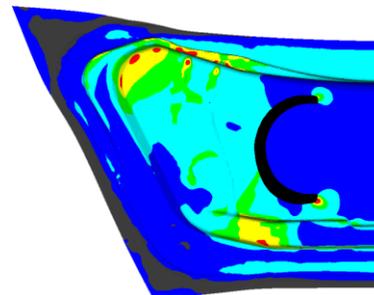


Product development & optimization

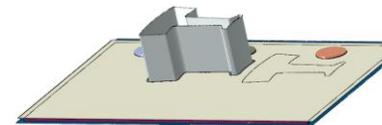


Simulations of technological processes

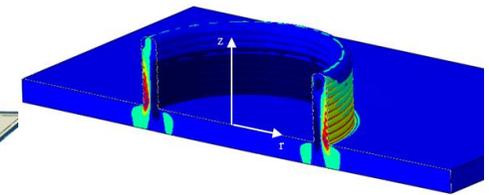
deep drawing, stamping



blanking



3D printing



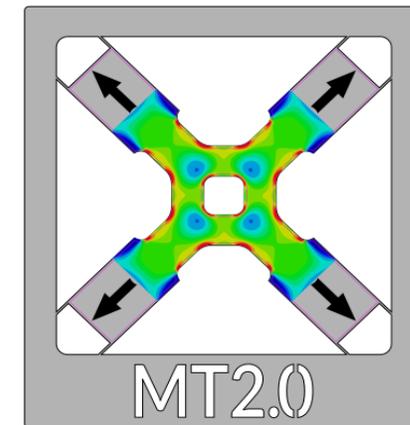
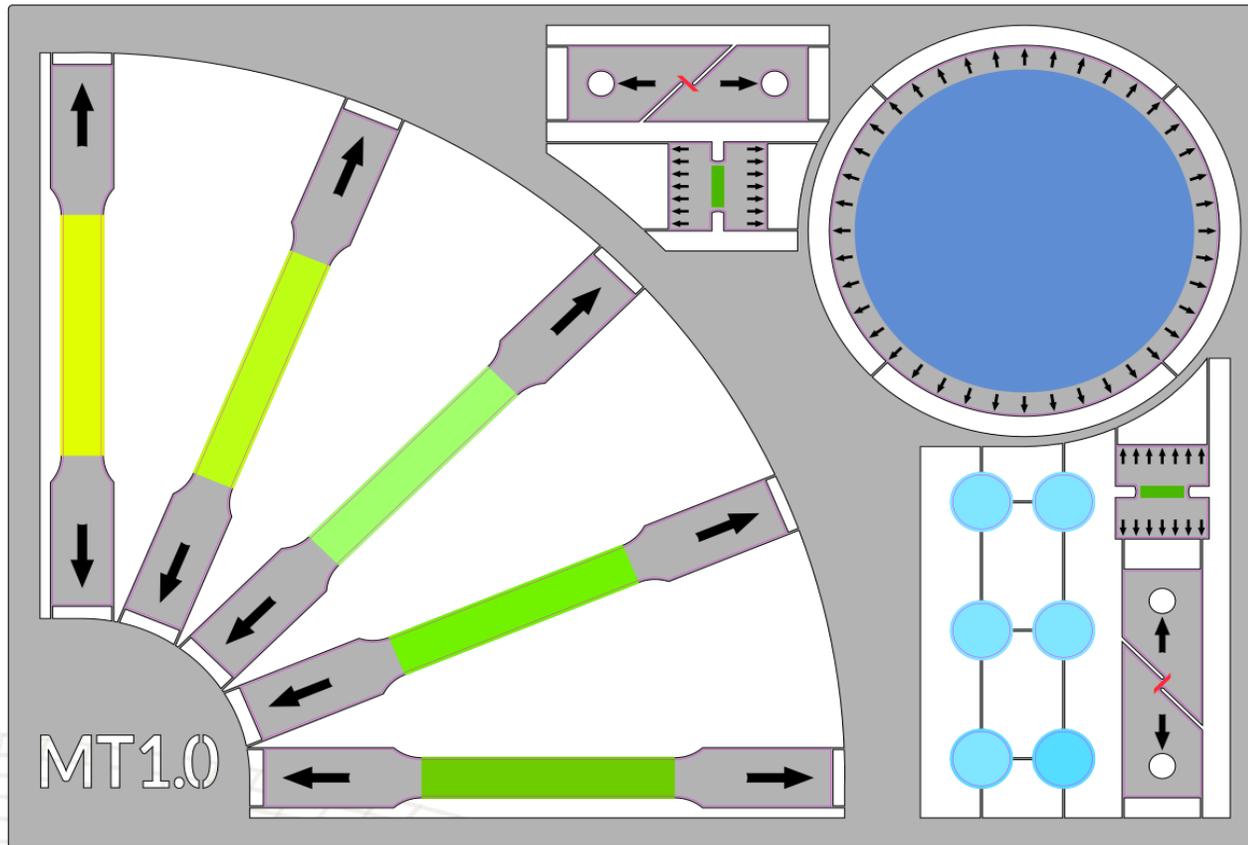
A history of MT 2.0...



MT 1.0 vs MT 2.0 in a nutshell

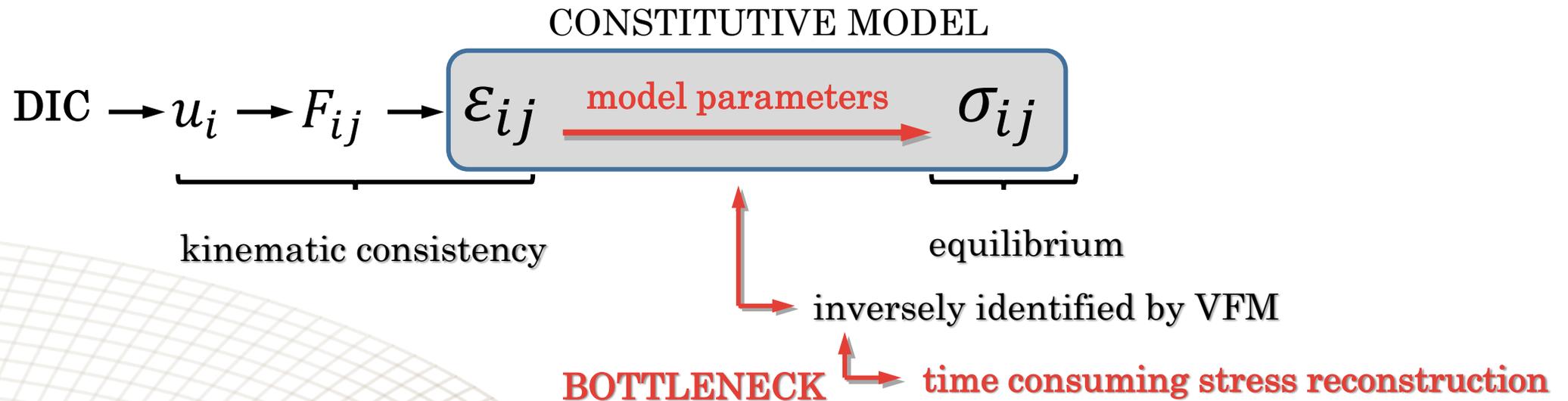
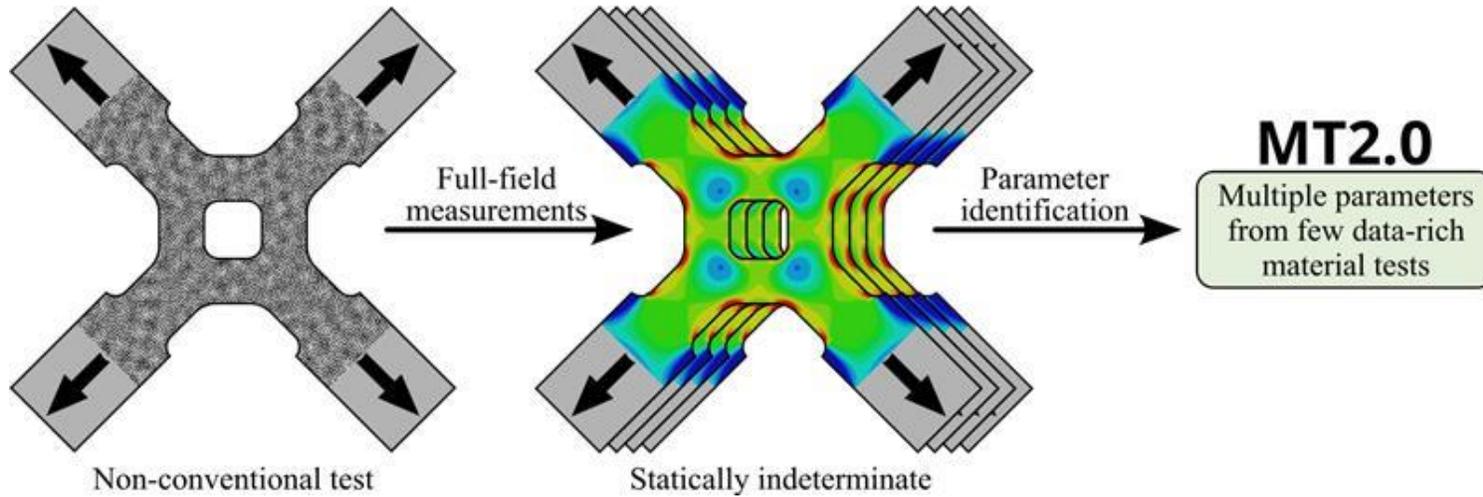


Material testing 1.0 (MT 1.0) vs Material testing 2.0 (MT 2.0):



Making the material characterization efficient, reliable and robust!

Making the material characterization efficient, reliable and robust!



- Backward Euler (BE)
 - implicit integration scheme
 - consistency condition fulfilled
 - stable
 - solution of system of nonlinear algebraic equations is required
 - iteration procedure in each increment is needed
 - difficult implementation of complex constitutive models
- Next Increment Corrects Error (NICE)
 - explicit integration scheme
 - no drifting from the consistency condition
 - stable
 - fast, efficient
 - simple implementation

NICE: an explicit numerical scheme for integration of constitutive equations, which efficiently combine the implementation simplicity of the forward-Euler scheme and accuracy of the backward-Euler scheme

A class of elastic-plastic constitutive models (system of DAE):

algebraic constraint/equation:

$$\Phi = \Phi(\sigma_{ij}, \sigma_Y, \kappa_1, \mathbf{K}, \kappa_p) = 0 \quad (\text{consistency condition})$$

differential (evolution) equations:

$$d\sigma_{ij} = C_{ijkl} d\varepsilon_{kl}^e = C_{ijkl} (d\varepsilon_{kl} - d\varepsilon_{kl}^p) \quad (\text{Hook's law})$$

$$d\varepsilon_{ij}^p = \frac{\partial \psi}{\partial \sigma_{ij}} d\lambda \quad (\text{flow rule})$$

$$\sigma_Y d\varepsilon_{eq}^p = \sigma_{ij} d\varepsilon_{ij}^p \quad (\text{uniaxial-multiaxial plastic work equivalence})$$

$$d\kappa_r = d\kappa_r(\sigma_{ij}, \sigma_Y, \kappa_1, \mathbf{K}, \kappa_p, d\varepsilon_{ij}^p); \quad (\text{additional state variables are supposed to follow specified differential laws})$$

$$r \in \{1, 2, \mathbf{K}, p\}$$

+ loading/unloading conditions

NICE (Next Increment Corrects Error) scheme:

How can we fulfill the algebraic constraint with explicit scheme?

algebraic constraint/equation:

$$\Phi = \Phi(\Sigma) = 0$$

Taylor power series expansion of the consistency condition on all unknown state variables

$$\Phi_n + \Delta\Phi = 0$$

differential (evolution) equations:

$$\Delta\Sigma = \mathbf{C} \cdot \Delta\varepsilon - \mathbf{R} \Delta\lambda$$

treated according to the forward-Euler approach

$$\Delta\Sigma = \mathbf{C} \cdot \Delta\varepsilon - \mathbf{R}_n \Delta\lambda$$

NICE (Next Increment Corrects Error) scheme:

NICE scheme:

$$\Phi_n + \Delta\Phi = \Phi_n + \left. \frac{\partial\Phi}{\partial\Sigma} \right|_n \cdot \Delta\Sigma = 0$$

$$\Delta\Sigma = \mathbf{C} \cdot \Delta\varepsilon - \mathbf{R}_n \Delta\lambda$$

explicit expression of
plastic multiplier

$$\Delta\lambda = \frac{\Phi_n + \left. \frac{\partial\Phi}{\partial\Sigma} \right|_n \cdot \mathbf{C} \cdot \Delta\varepsilon}{\left. \frac{\partial\Phi}{\partial\Sigma} \right|_n \cdot \mathbf{R}_n}$$

Forward-Euler:

$$\Delta\Phi = \left. \frac{\partial\Phi}{\partial\Sigma} \right|_n \cdot \Delta\Sigma = 0$$

$$\Delta\Sigma = \mathbf{C} \cdot \Delta\varepsilon - \mathbf{R}_n \Delta\lambda$$

explicit expression of
plastic multiplier

$$\Delta\lambda = \frac{\left. \frac{\partial\Phi}{\partial\Sigma} \right|_n \cdot \mathbf{C} \cdot \Delta\varepsilon}{\left. \frac{\partial\Phi}{\partial\Sigma} \right|_n \cdot \mathbf{R}_n}$$

Simple mathematical example: 2-D loading of membrane

algebraic constraint/equation:

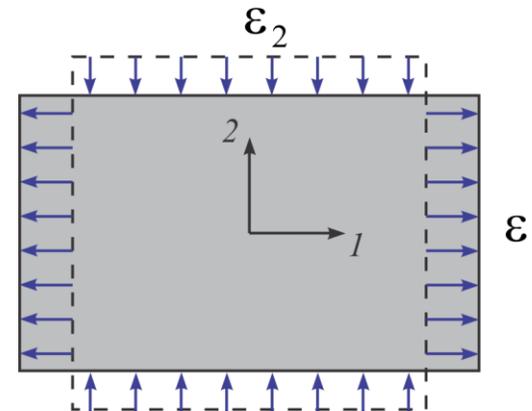
$$\Phi(\sigma_1, \sigma_2, \sigma_Y) = \frac{\sigma_1^2 + \sigma_2^2}{\sigma_Y^2} - 1 = 0$$

differential (evolution) equations:

$$d\sigma_1 = E d\varepsilon_1 - E \frac{\partial \Phi}{\partial \sigma_1} d\lambda$$

$$d\sigma_2 = E d\varepsilon_2 - E \frac{\partial \Phi}{\partial \sigma_2} d\lambda$$

$$d\sigma_Y = H d\lambda$$



Example: loading of membrane element



Simple mathematical example: 2-D loading of membrane

algebraic constraint/equation:

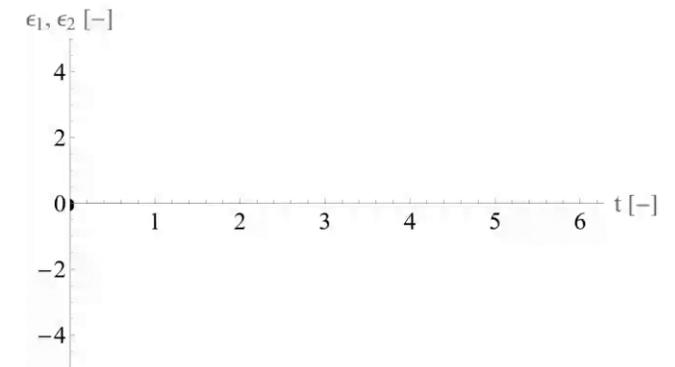
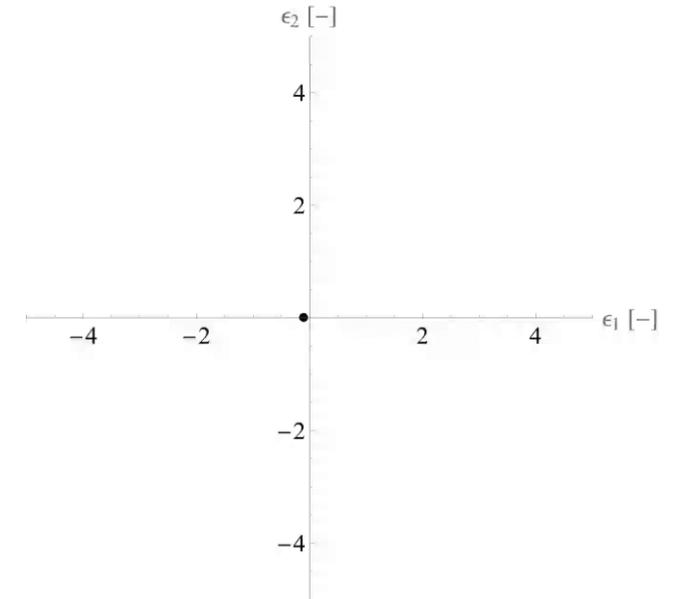
$$\Phi_n + \Delta\Phi = \Phi_n + \left. \frac{\partial\Phi}{\partial\Sigma} \right|_n \cdot \Delta\Sigma = 0 \quad \longrightarrow$$

$$\Delta\lambda = \frac{\Phi_n + \left. \frac{\partial\Phi}{\partial\Sigma} \right|_n \cdot \mathbf{C} \cdot \Delta\boldsymbol{\varepsilon}}{\left. \frac{\partial\Phi}{\partial\Sigma} \right|_n \cdot \mathbf{R}_n}$$

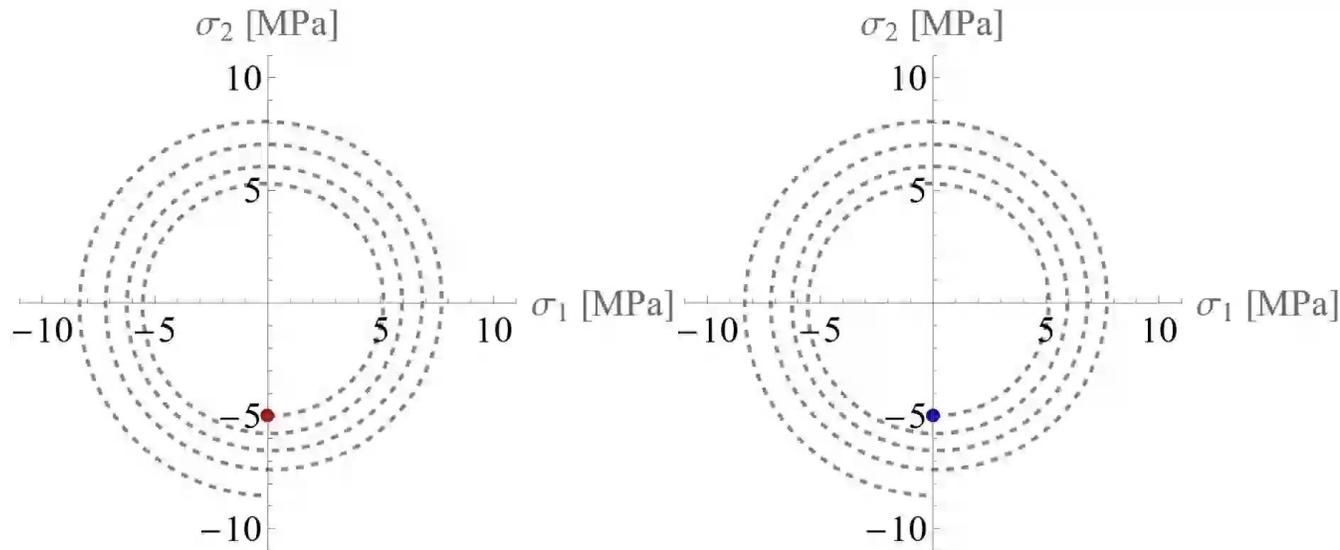
differential (evolution) equations:

$$\Delta\Sigma = \mathbf{C} \cdot \Delta\boldsymbol{\varepsilon} - \mathbf{R}_n \Delta\lambda$$

$$\Delta\boldsymbol{\varepsilon} = \begin{Bmatrix} \Delta\varepsilon_1 \\ \Delta\varepsilon_2 \end{Bmatrix}; \Delta\Sigma = \begin{Bmatrix} \Delta\sigma_1 \\ \Delta\sigma_2 \\ \Delta\sigma_Y \end{Bmatrix}; \mathbf{C} = \begin{bmatrix} E & 0 \\ 0 & E \\ 0 & 0 \end{bmatrix}; \mathbf{R} = \begin{Bmatrix} 2E\sigma_1\sigma_Y^{-2} \\ 2E\sigma_2\sigma_Y^{-2} \\ -H \end{Bmatrix}$$



Simple mathematical example: 2-D loading of membrane



NICE

Forward Euler

Next Increment Corrects Error (NICE)

- explicit integration scheme
- no drifting from the consistency condition
- stable

- fast, efficient
- simple implementation
- ✗ substepping for large increments since the error can get large (implicit FEM)

Boosting the efficiency of non-linear VFM by implementation of NICE

NICE

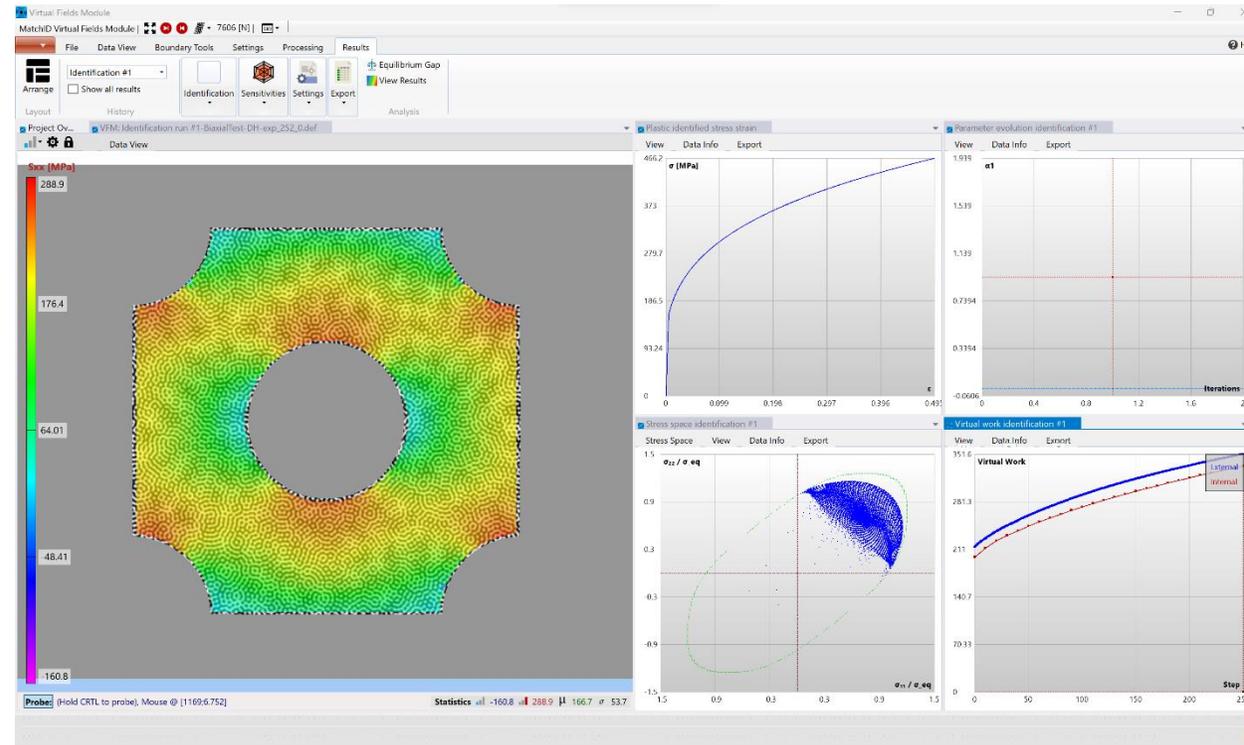


MatchID

Metrology beyond colors

The Virtual Fields
Method

Extracting Constitutive Mechanical
Parameters from Full-field Deformation
Measurements



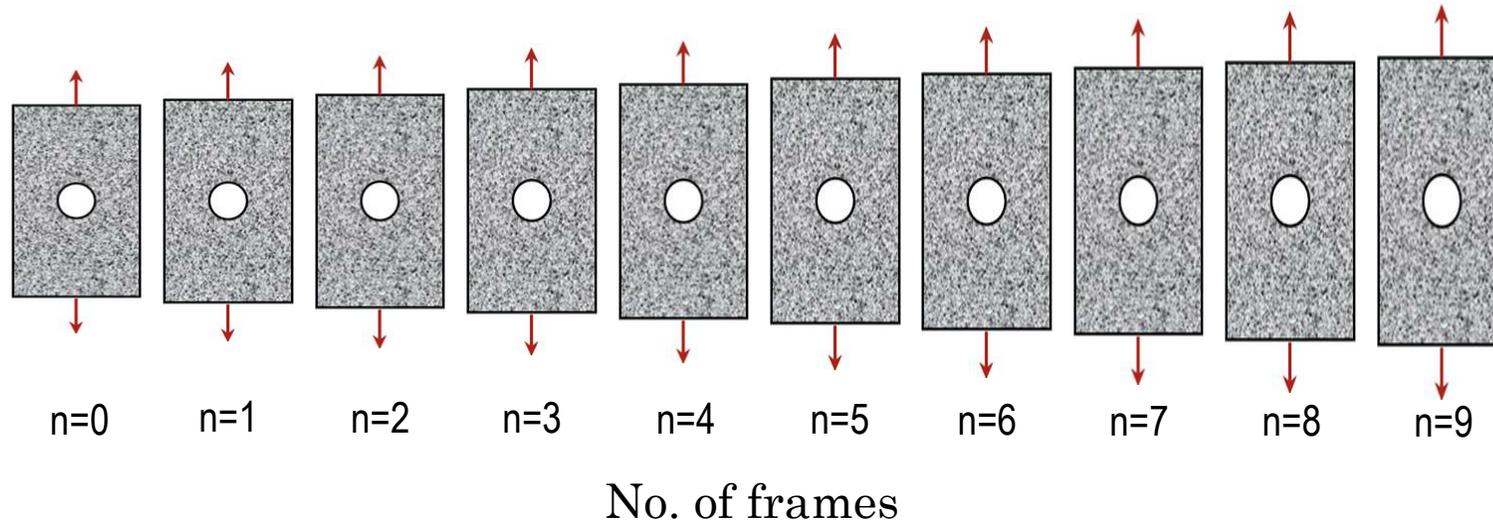
user interface: measure and postprocess full-field test data, apply boundary cond., material model, choose parameters, optimization algorithm, initial conditions, few details and run!

Effect of temporal down-sampling



Image acquisition for DIC

Linear elasticity



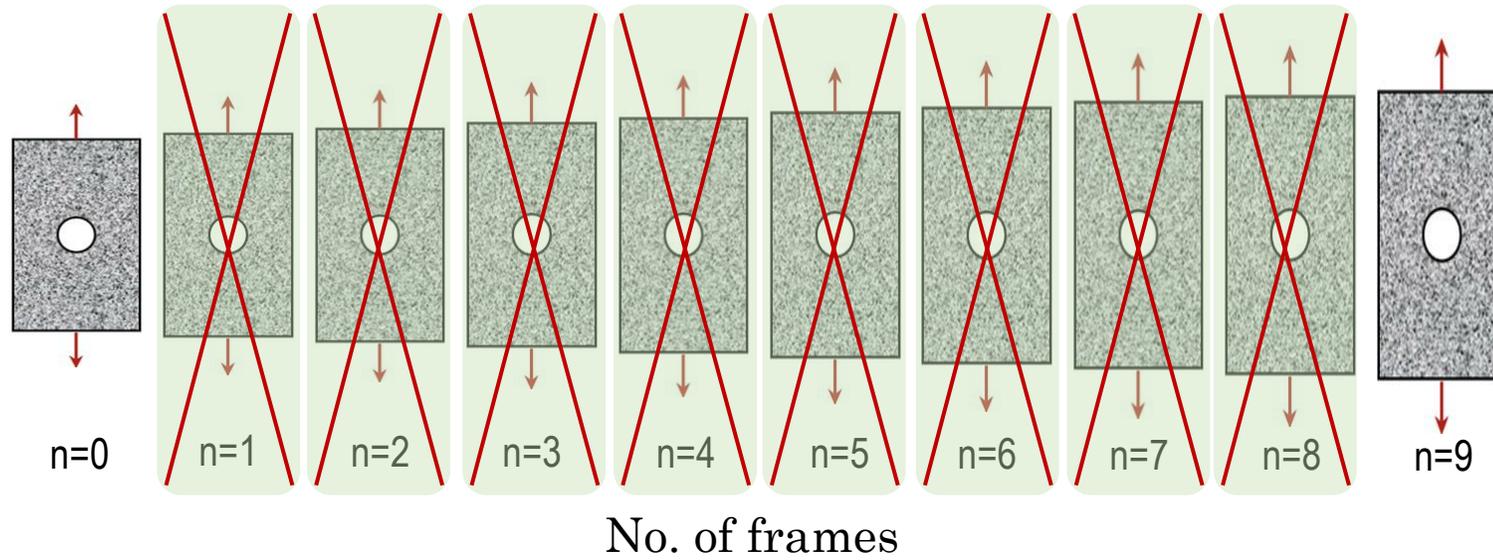
The user has control on frame acquisition rate \longrightarrow improved efficiency?

Effect of temporal down-sampling



Image acquisition for DIC

Linear elasticity



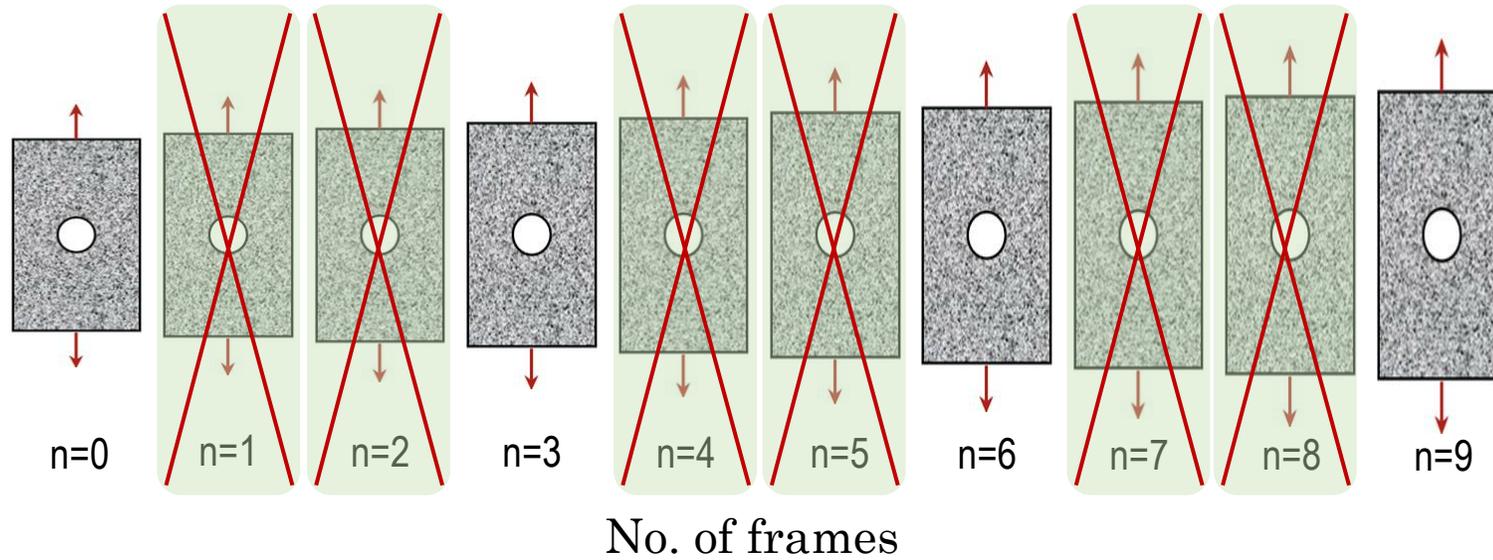
The user has control on frame acquisition rate \longrightarrow improved efficiency?

Effect of temporal down-sampling



Image acquisition for DIC

Linear elasticity ✓

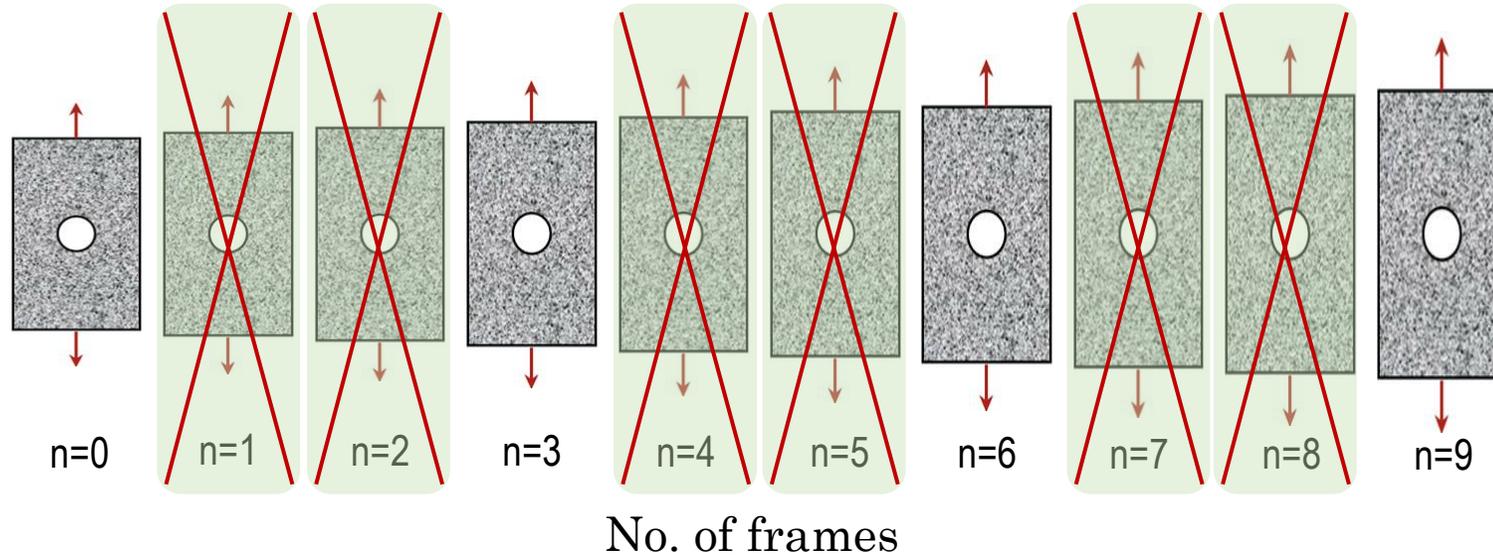


The user has control on frame acquisition rate → improved efficiency?

Effect of temporal down-sampling



Image acquisition for DIC



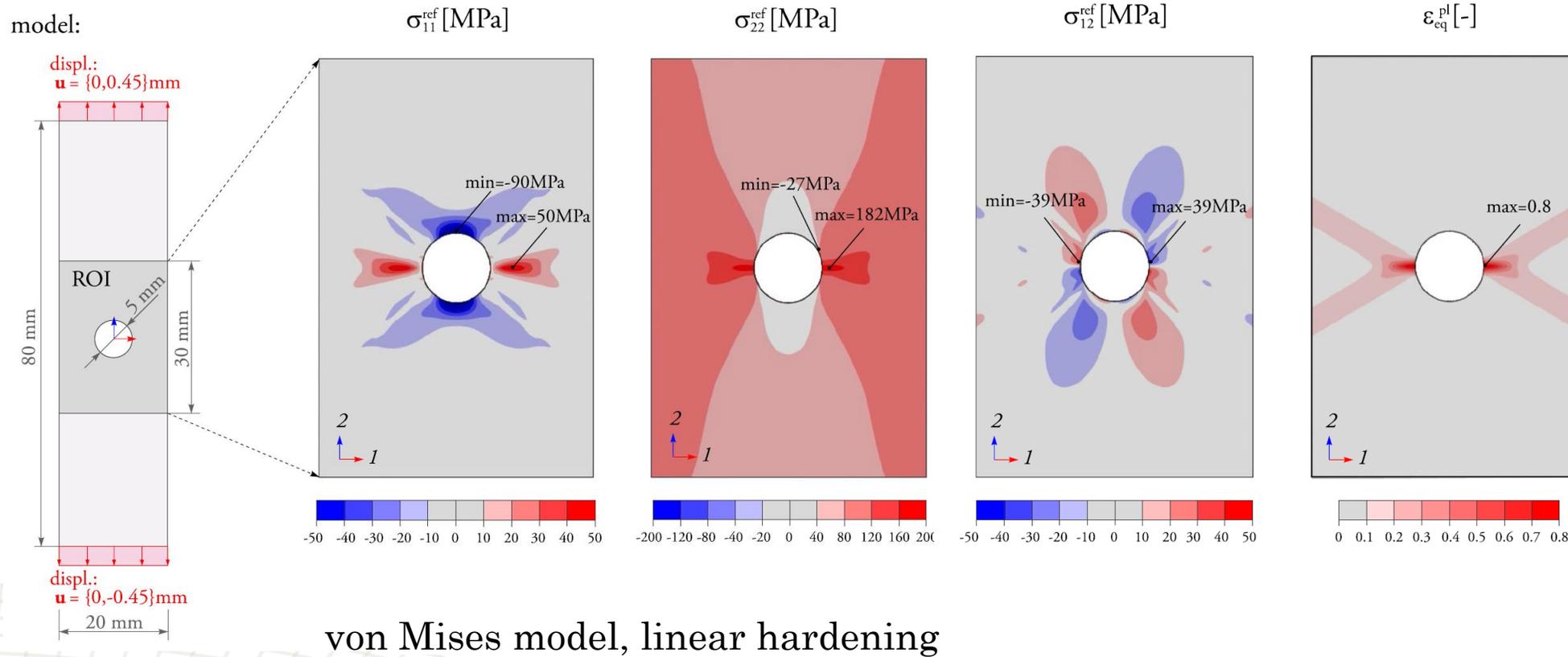
Linear elasticity ✓

Plasticity ✗

↙ results: path dependent!

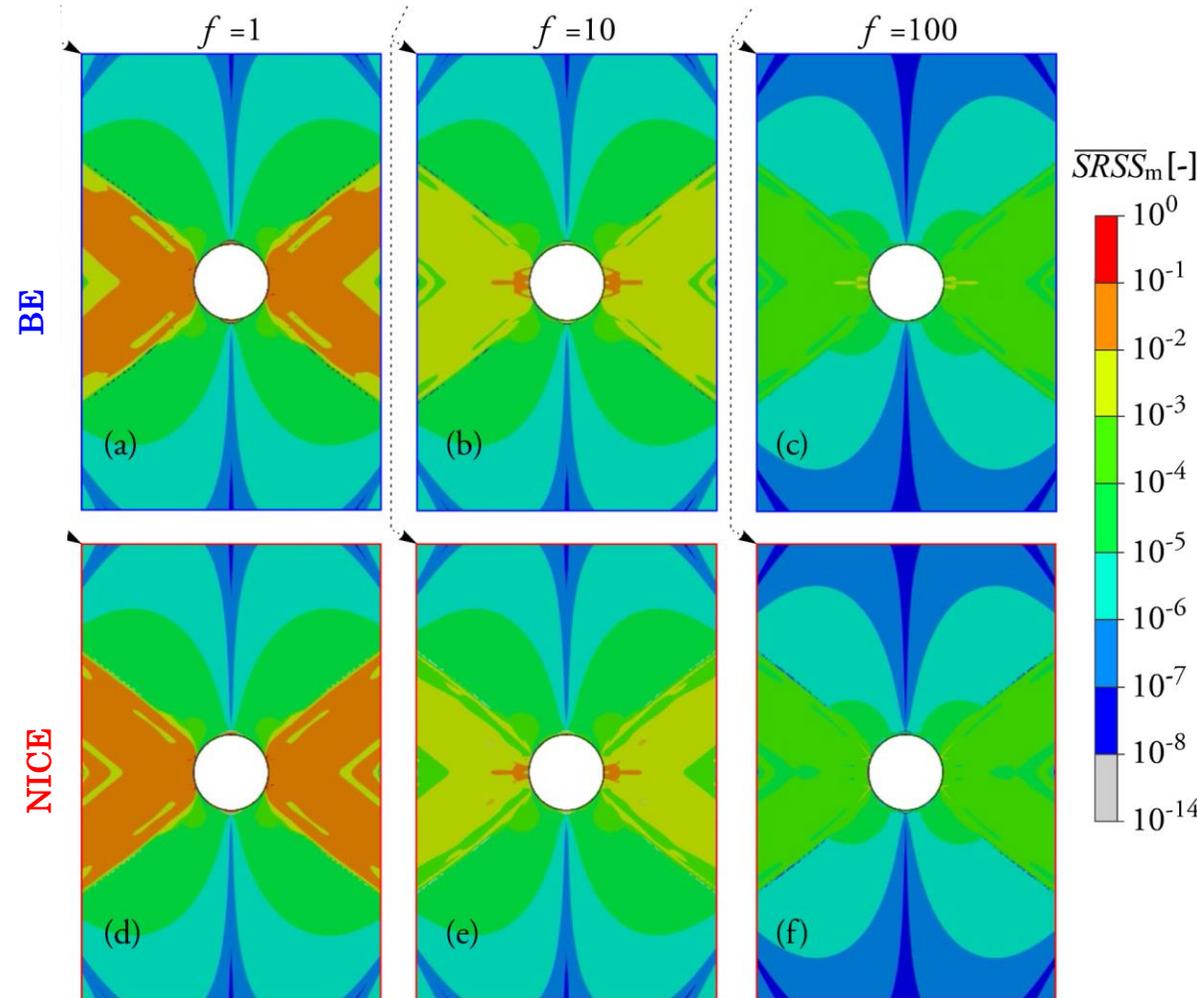
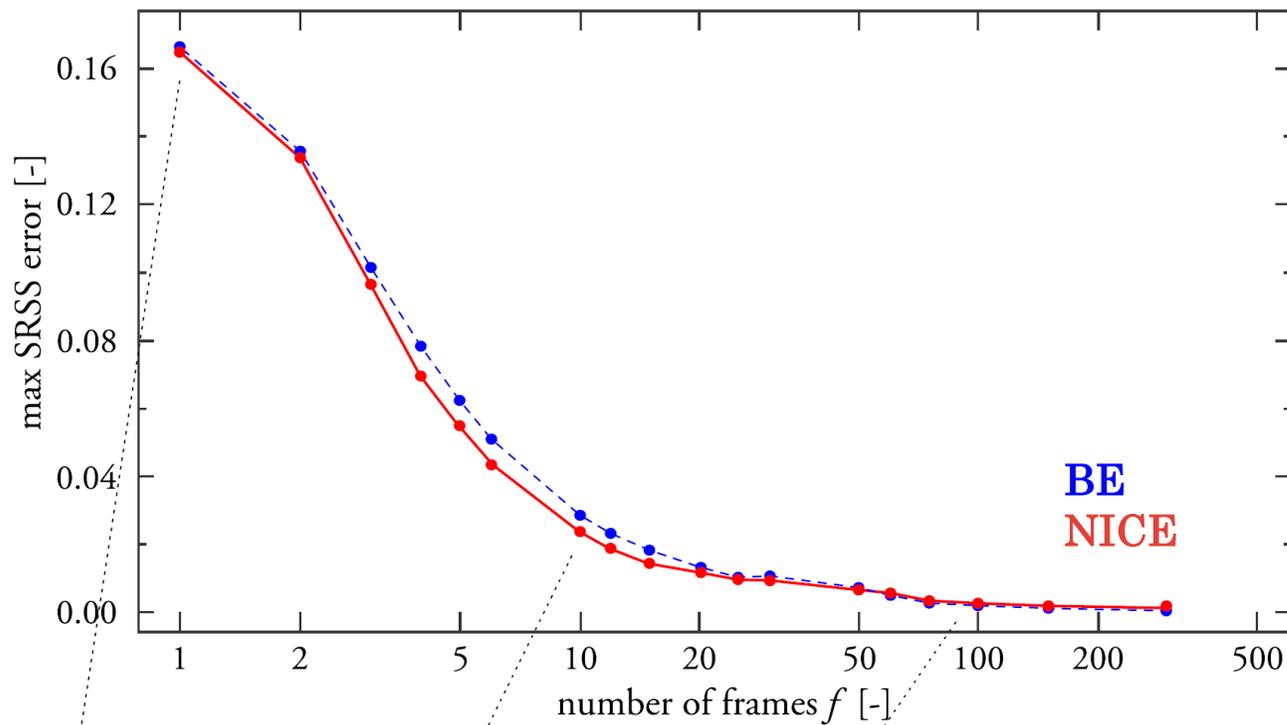
The user has control on frame acquisition rate → improved efficiency?

FEA results for the open hole test

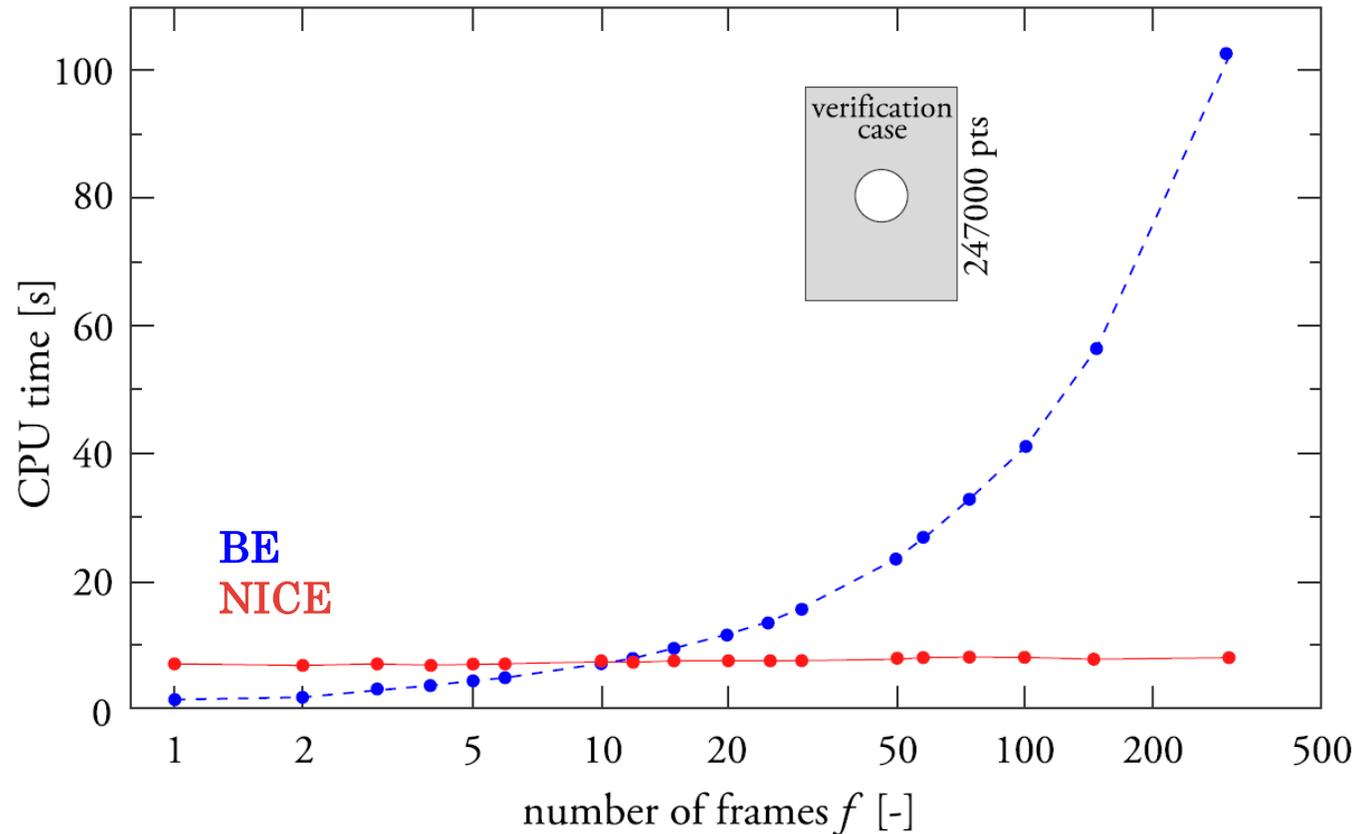


FEA results for the open hole test

Equidistant incrementation

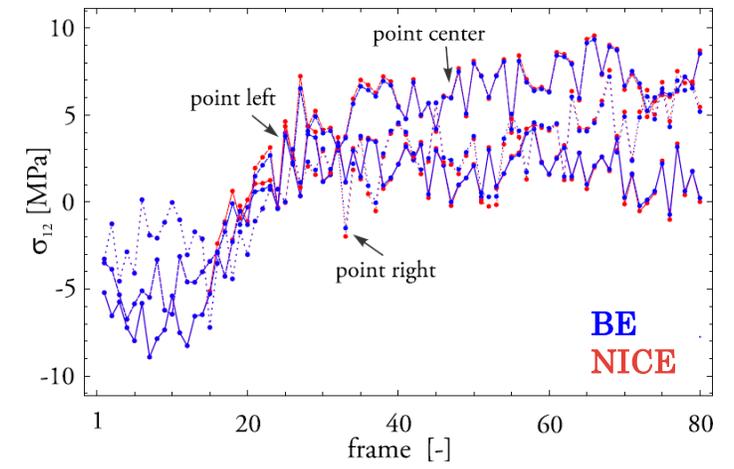
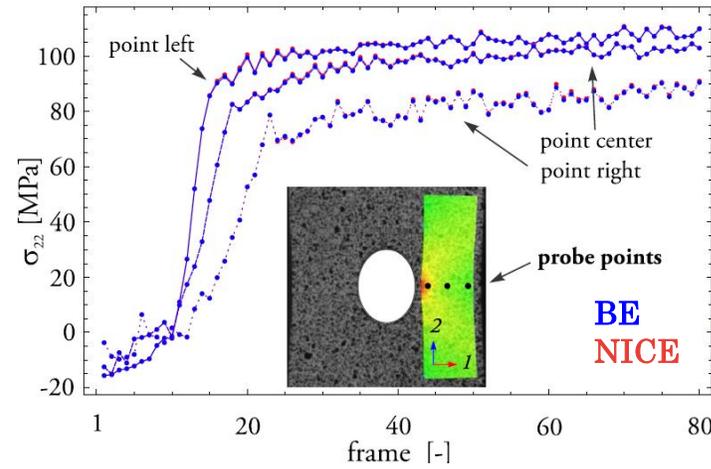
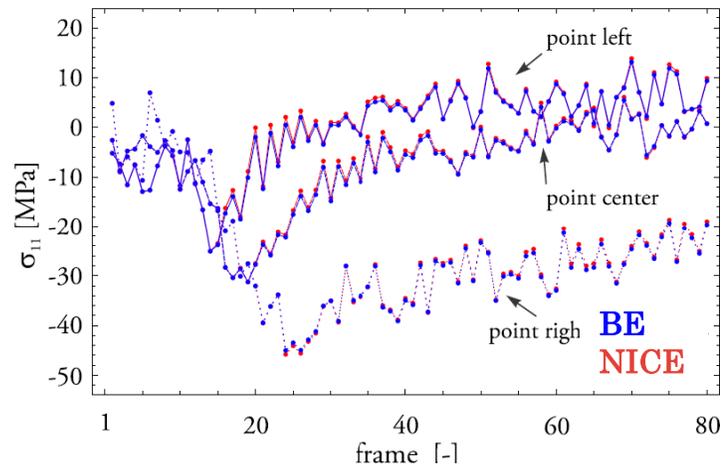


Stress reconstruction efficiency



Theoretically: using NICE, there is no need for down-sampling to save CPU time.

The effect of strain reconstruction on stress integration



GOAL: Implementation of NICE into MatchID via CMAT module

- identification of plastic anisotropy using YLD2000-2d model parameters using perforated biaxial test
 - monitor efficiency, accuracy and robustness
- Material model: Swift hardening + YLD2000-2d
 - Identification of material parameters:
 - $\alpha_1, \alpha_2, \alpha_3, \alpha_4, \alpha_5, \alpha_6, \alpha_7, \alpha_8, m$
 - Virtual experiment via biaxial test

FEM model and simulation with a known material parameters

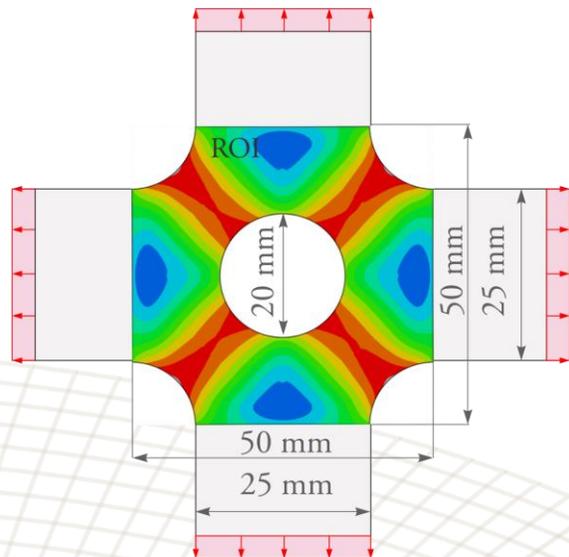
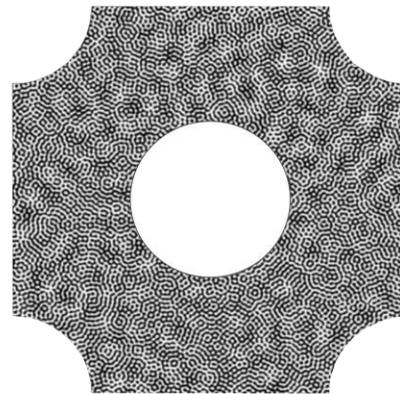
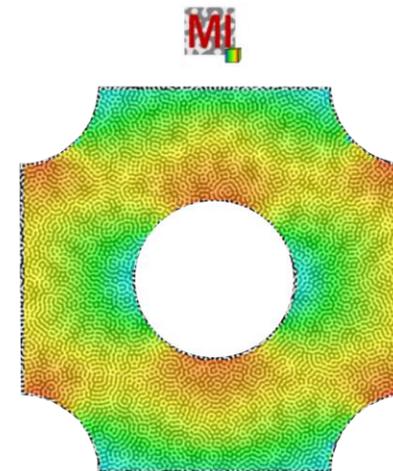


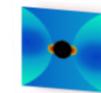
Image deformation with synthetic speckle pattern (no noise)

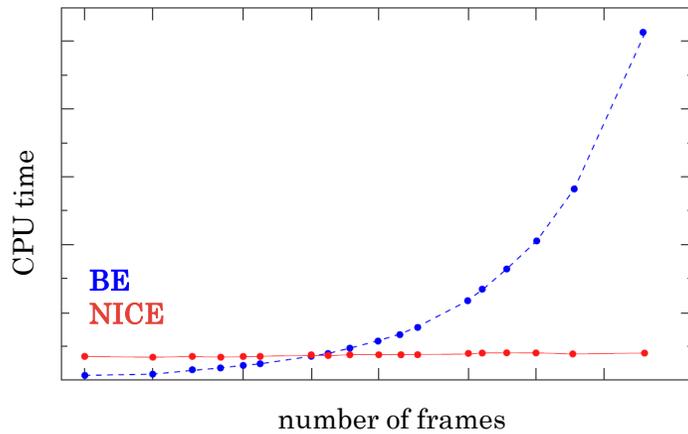
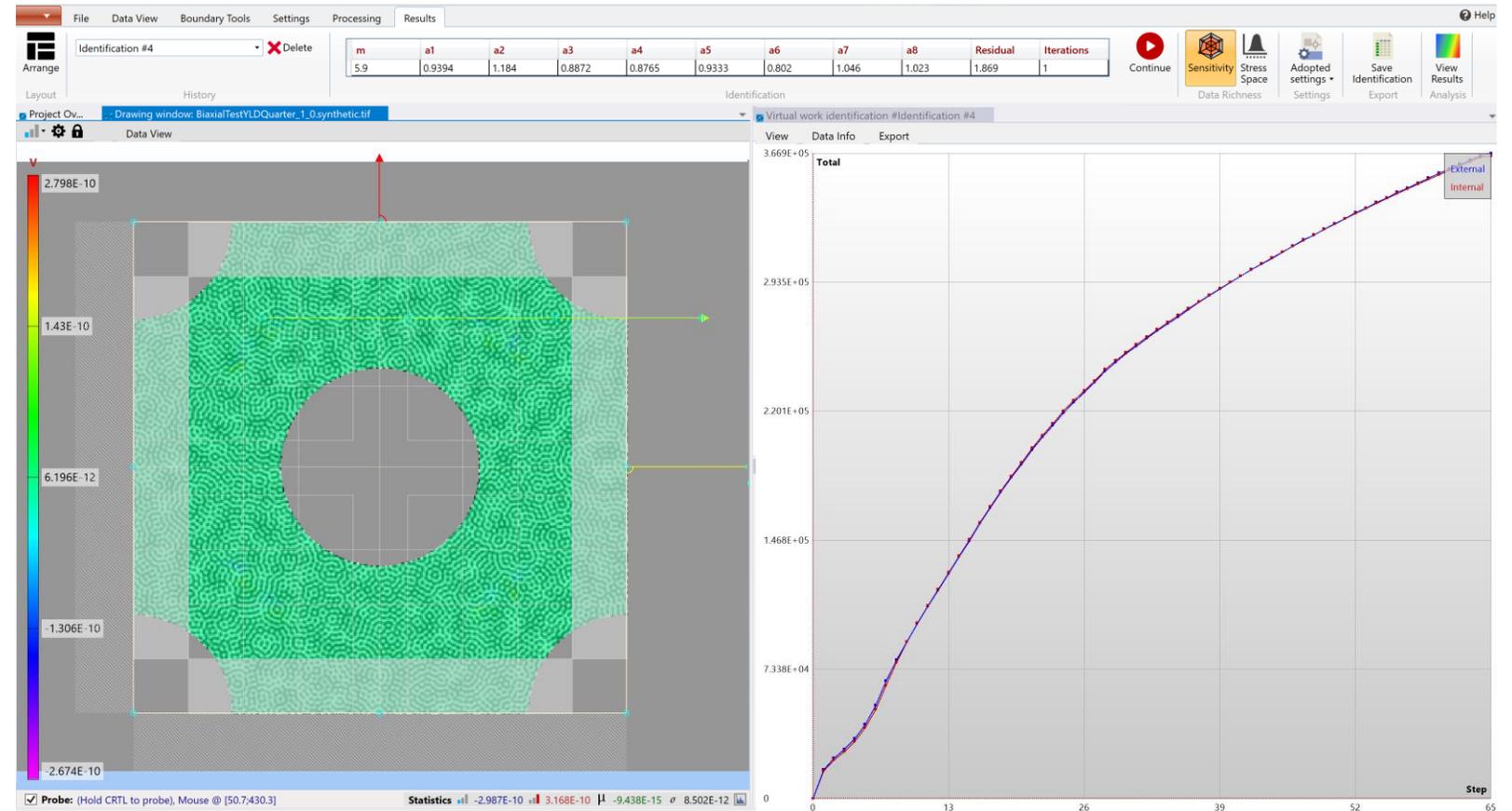


DIC processing of deformed synthetic images



Inverse identification of parameters via nonlinear VFM





Efficiency: the NICE algorithm is faster in comparison to BE, depending on a number of frames, parameters and iterations

Findings:

- In plasticity, temporal down-sampling of the DIC results increases inaccuracy of results
- The NICE method enables efficient stress reconstruction for highly nonlinear problems
- Using NICE, there is no reason to perform down-sampling to save computational time
- The NICE algorithm was successfully implemented into MatchID via CMAT, and proven to be computationally efficient
- Using NICE computational efficiency can be increased → **larger the number of stress reconstructions, larger the benefit of NICE against BE**

More info:

Halilovič, Starman, Coppeters: Computationally Efficient Stress Reconstruction from Full-field Strain Measurements, Computational Mechanics, 2024

Thank you for your attention