

VERS UNE APPROCHE NON-PARAMÉTRIQUE EN MÉCANIQUE DES MATÉRIAUX

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UNIVERSITÉ DE NANTES



CONTINUUM MECHANICS

Find $(\mathbf{u}, \boldsymbol{\sigma})$ such that:

- Compatibility:

$$\boldsymbol{\varepsilon} = \frac{1}{2}(\nabla \mathbf{u} + \nabla^T \mathbf{u})$$

- Constitutive relation:

$$\boldsymbol{\sigma} = f(\boldsymbol{\varepsilon}, \dots)$$

- Balance of momentum

$$\operatorname{div}(\boldsymbol{\sigma}) = 0$$

- Boundary conditions

$$\mathbf{u} = \mathbf{u}_d \text{ on } \Gamma_u$$

$$\boldsymbol{\sigma}(\mathbf{n}) = \mathbf{F}_d \text{ on } \Gamma_F$$

CONTINUUM MECHANICS

Find $(\mathbf{u}, \boldsymbol{\sigma})$ such that:

- Admissibility for \mathbf{u} :

$$\mathbf{u} = \mathbf{u}_d \text{ on } \Gamma_u + \text{regularity}$$

- Admissibility for $\boldsymbol{\sigma}$:

$$\operatorname{div}(\boldsymbol{\sigma}) = 0$$

$$\boldsymbol{\sigma}(\mathbf{n}) = \mathbf{F}_d \text{ on } \Gamma_F$$

- Constitutive relation:

$$\boldsymbol{\sigma} = f(\boldsymbol{\varepsilon}, \dots)$$

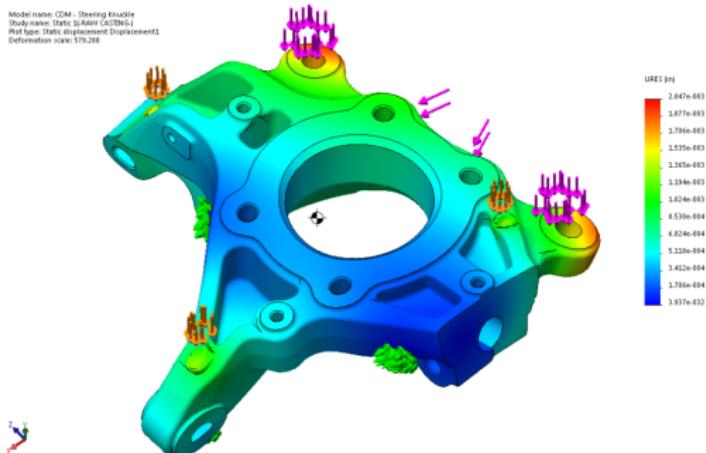
with $\boldsymbol{\varepsilon} = \frac{1}{2}(\nabla \mathbf{u} + \nabla^T \mathbf{u})$

CONTINUUM MECHANICS

Resolution:

Minimization of an energy functional / Principle of Virtual Work

under constraint: $\sigma = f(\varepsilon, \dots)$



LIMITATIONS OF CONSTITUTIVE LAWS

Complex Materials:

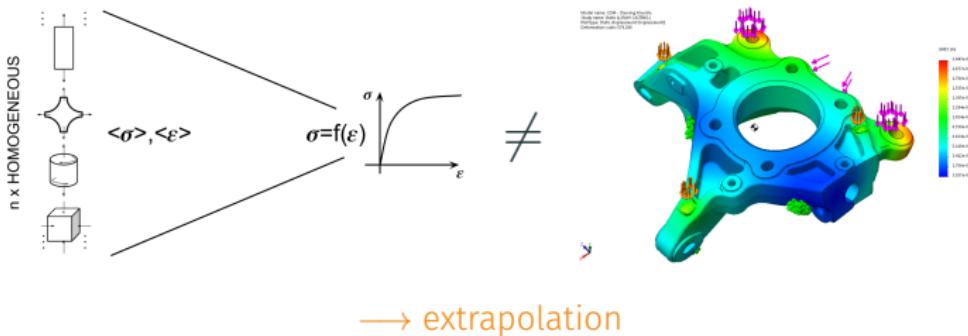
- continuum inherits all the complexity of the smaller scales

Advanced theoretical framework:

- non-linear, history dependent, ...
- $\boldsymbol{\varepsilon}, \boldsymbol{\sigma}$ are 2nd order tensors

Too simple experiments:

- homogeneous $\boldsymbol{\varepsilon}, \boldsymbol{\sigma}$
- mainly uniaxial



CONTINUUM MECHANICS

Find $(\mathbf{u}, \boldsymbol{\sigma})$ such that:

- Admissibility for \mathbf{u} :

$$\mathbf{u} = \mathbf{u}_d \text{ on } \Gamma_u + \text{regularity}$$

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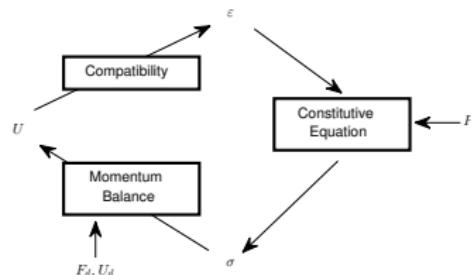
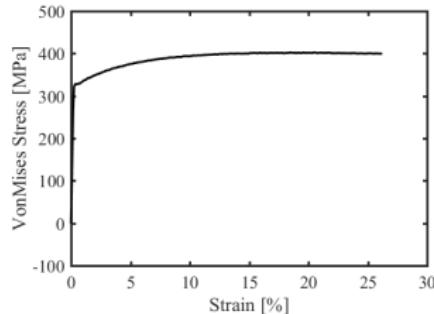
- Constitutive relation:

$$\boldsymbol{\sigma} = f(\boldsymbol{\varepsilon}, \dots)$$

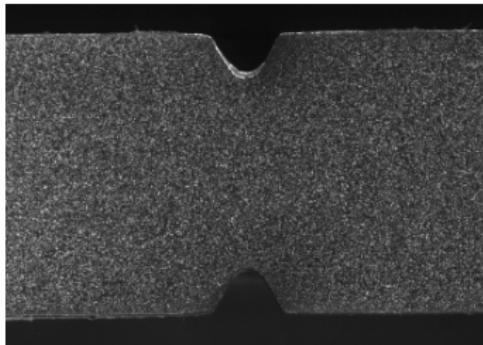
with $\boldsymbol{\varepsilon} = \frac{1}{2}(\nabla \mathbf{u} + \nabla^T \mathbf{u})$

LIMITATIONS OF CONSTITUTIVE LAWS

- ▶ “Engineering” approach

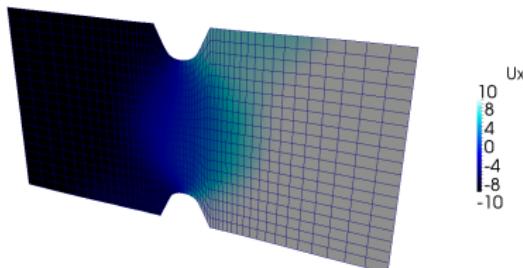
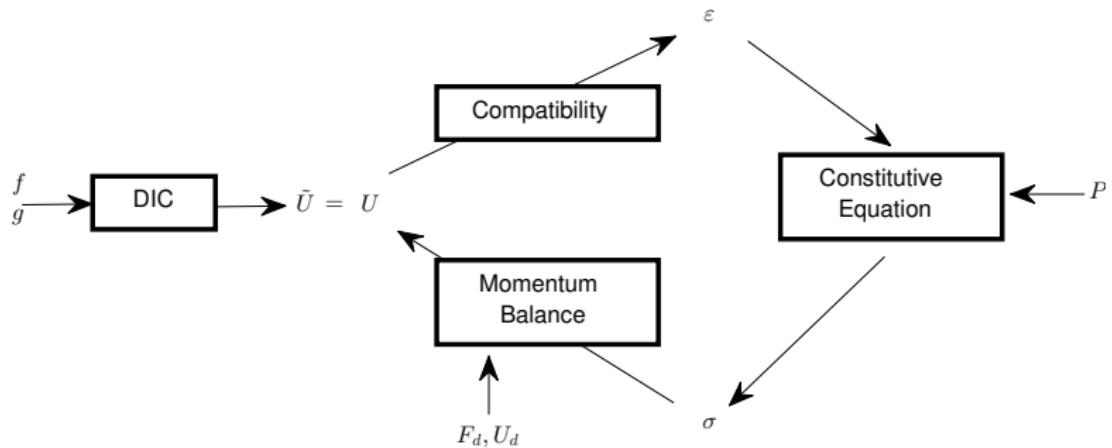


- ▶ Validation ?



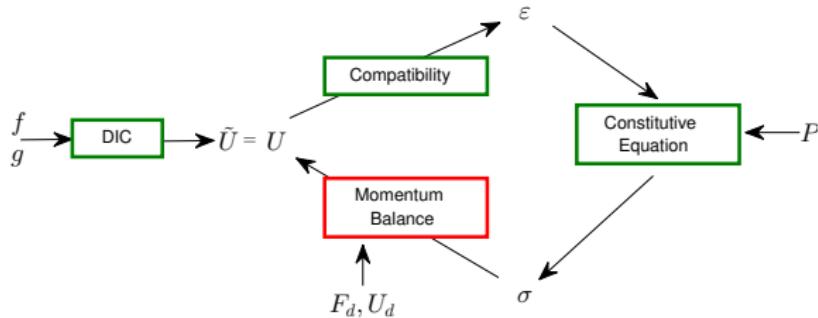
LIMITATIONS OF CONSTITUTIVE LAWS

► Photomechanics

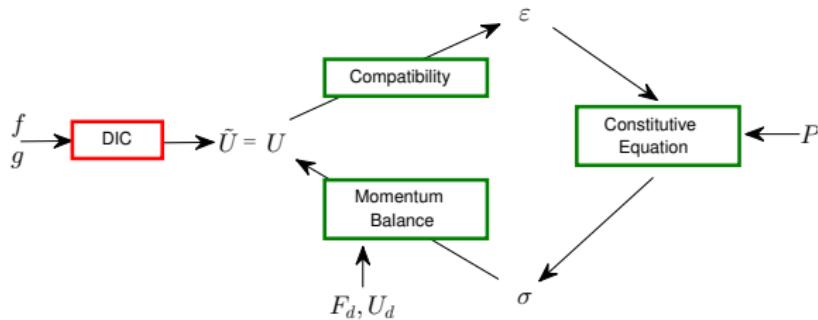


LIMITATIONS OF CONSTITUTIVE LAWS

► Stress calculation

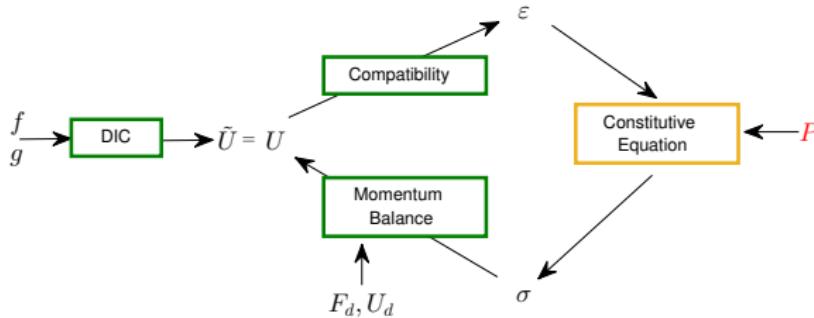


► Numerical simulation

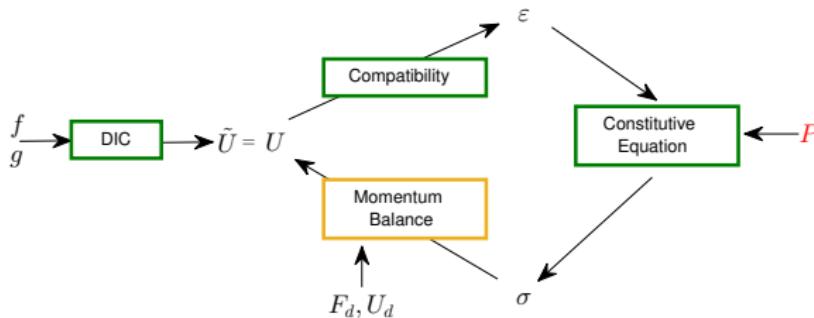


LIMITATIONS OF CONSTITUTIVE LAWS

- ▶ Constitutive Equation Gap [Chrysochoos *et al.*]

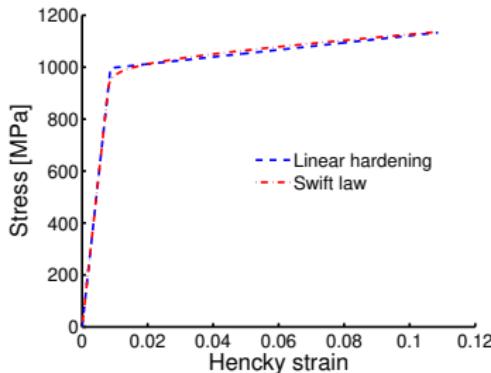
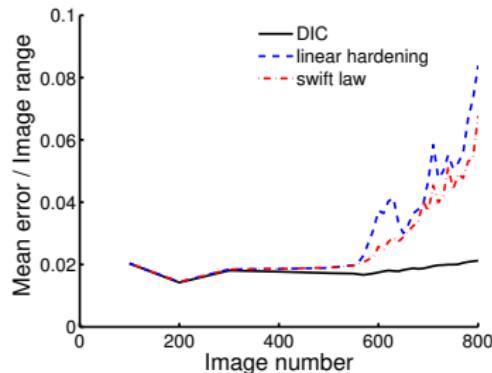
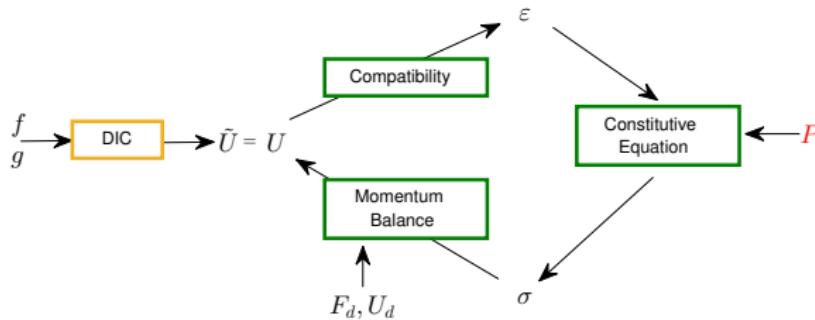


- ▶ Equilibrium gap [Hild *et al.*], VFM [Grédiac *et al.*]



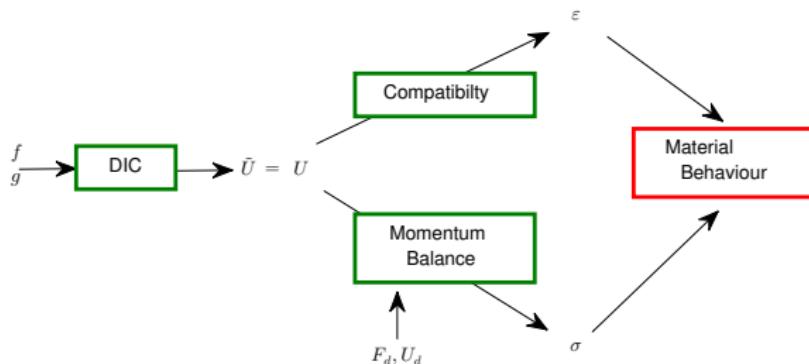
LIMITATIONS OF CONSTITUTIVE LAWS

- FEMU [Lecompte et al., Leclerc et al....]

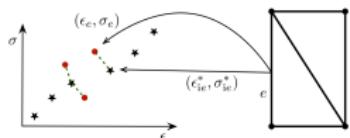


PARAMETRIC V.S. NON-PARAMETRIC

- ▶ Parametric techniques (using a constitutive equation)
 - + provide for the optimal set of parameters
 - +/- tell that the constitutive equation is not *correct*
 - how to improve it
 - kinematically consistent direct FEA
- ▶ Non-parametric (without using a constitutive equation)



TOWARDS A NEW PARADIGM USING DATA BEYOND LAWS



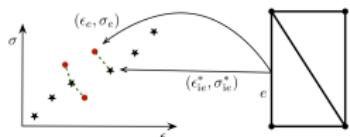
- ▶ use *data* beyond *laws*
- ▶ Min distance w.r. data under BE constraint
- ▶ Kirchdoerfer & Ortiz (2016)

data are obtained from *simple* experiments → limitations remain

Goals

- ▶ estimate heterogeneous material state fields
 - ▶ strain
 - ▶ stress
 - ▶ ...
- ▶ no pre-supposed constitutive equations

TOWARDS A NEW PARADIGM USING DATA BEYOND LAWS



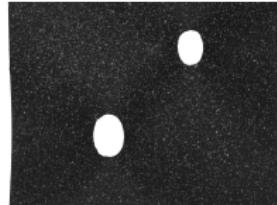
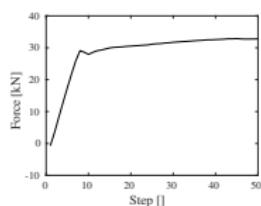
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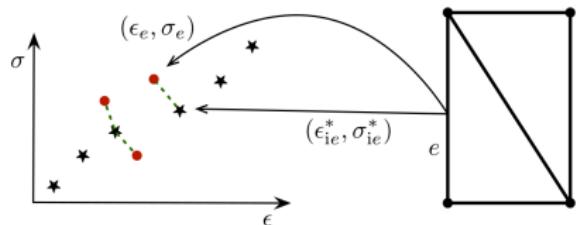
1. Inverse data-driven → non-linear elasticity



2. Mixed reduction → plasticity



DATA-DRIVEN COMPUTATIONAL MECHANICS



- $(\epsilon_{ie}^*, \sigma_{ie}^*)$ = material state
- (ϵ_e, σ_e) = mechanical state
- ie mapping

Minimization of the *distance* between (ϵ_e, σ_e) and $(\epsilon_{ie}^*, \sigma_{ie}^*)$

$$\text{solution} = \arg \min_{\epsilon_e, \sigma_e, ie} \frac{1}{2} \sum_e W_e \left(C_e (\epsilon_e - \epsilon_{ie}^*)^2 + \frac{1}{C_e} (\sigma_e - \sigma_{ie}^*)^2 \right)$$

subjected to:

- Balance of momentum

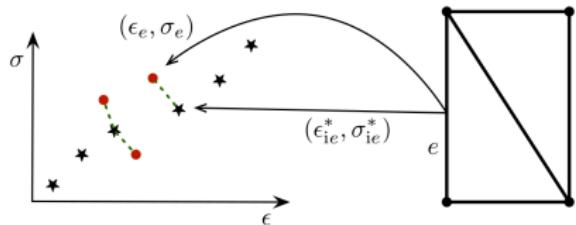
$$\sum_e W_e B_{ej} \sigma_e = f_j$$

- Strain compatibility

$$\epsilon_e = \sum_j B_{ej} u_j$$

written for truss lattice

DATA-DRIVEN IDENTIFICATION



- $(\epsilon_{ie}^*, \sigma_{ie}^*)$ = material state
- (ϵ_e, σ_e) = mechanical state
- ie mapping

Minimization of the *distance* between (ϵ_e, σ_e) and $(\epsilon_{ie}^*, \sigma_{ie}^*)$

$$\text{solution} = \arg \min_{\epsilon_e^*, \sigma_e^*, \sigma_e, ie} \frac{1}{2} \sum_e w_e \left(C_e (\epsilon_e - \epsilon_{ie}^*)^2 + \frac{1}{C_e} (\sigma_e - \sigma_{ie}^*)^2 \right)$$

subjected to:

- Balance of momentum

$$\sum_e w_e B_{ej} \sigma_e = f_j$$

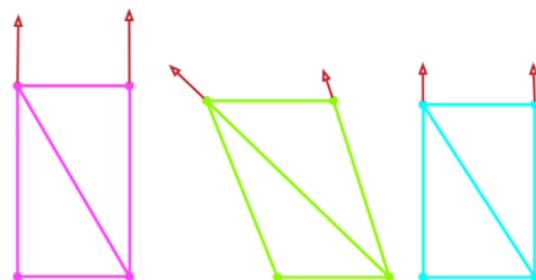
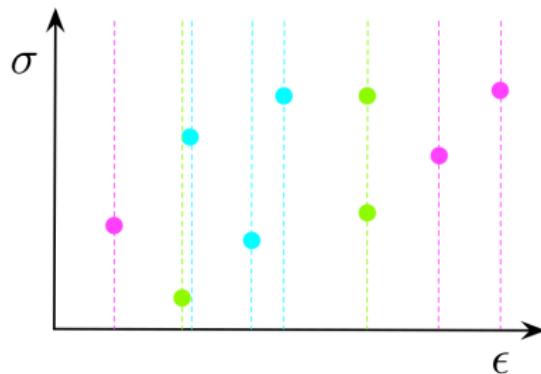
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How IT WORKS....

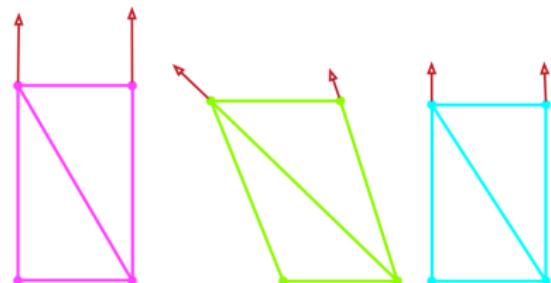
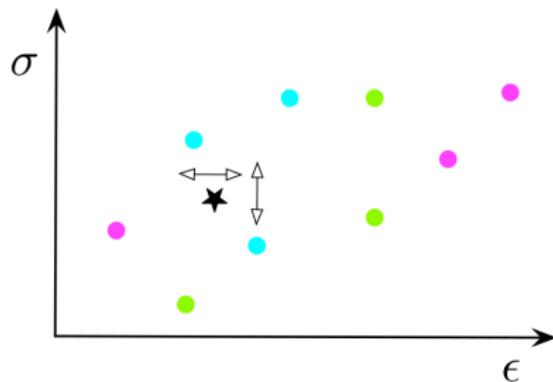
Starting from strain measurements



the mechanical points can only move vertically

How IT WORKS....

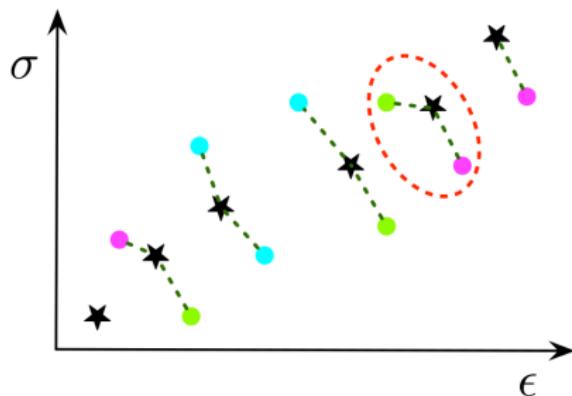
Starting from strain measurements



but the material states are unknown

HOW IT WORKS....

Starting from strain measurements



$$\min_{\epsilon_i^*, \sigma_i^*, \sigma_e, ie} \sum_{\text{snapshots}} \sum_e \left(C_e (\epsilon_e - \epsilon_{ie}^*)^2 + \frac{1}{C_e} (\sigma_e - \sigma_{ie}^*)^2 \right)$$

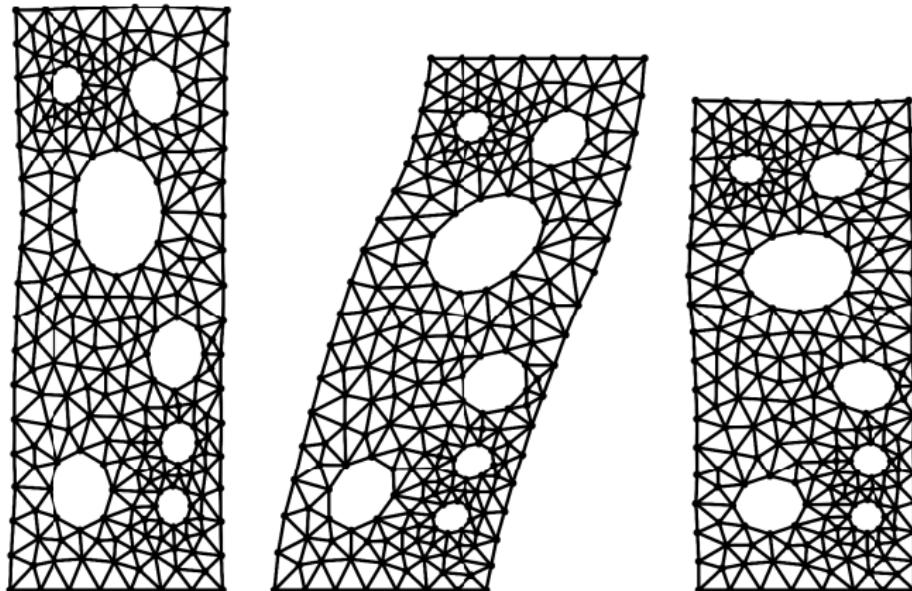
↓ ↓ ↓

* ● *

The diagram illustrates the optimization process. It starts with a set of strain measurements (stars) and stress measurements (circles) plotted against strain. These are used to calculate optimal values for ϵ_i^* , σ_i^* , and σ_e . These optimal values are then used to calculate the error for each snapshot e .

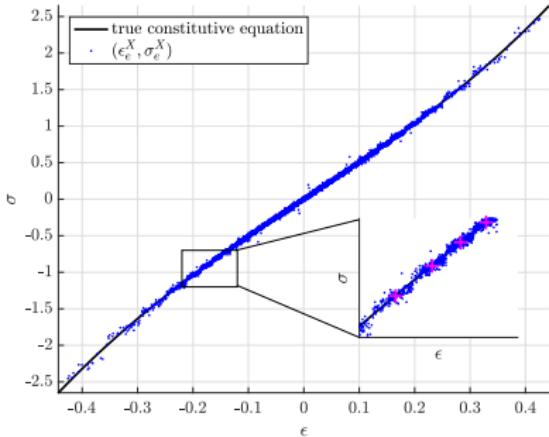
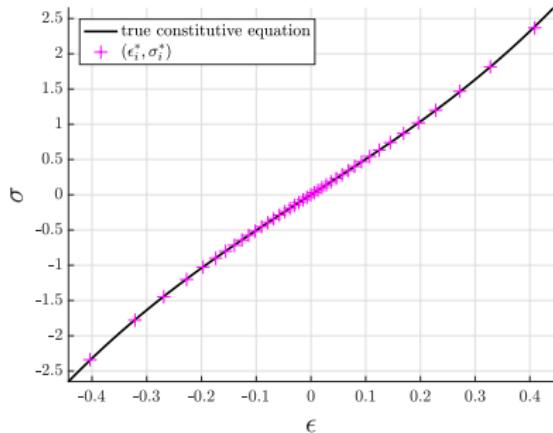
HOW IT WORKS....

Example (3 snapshots, 40 material states, 650 mechanical states)



HOW IT WORKS....

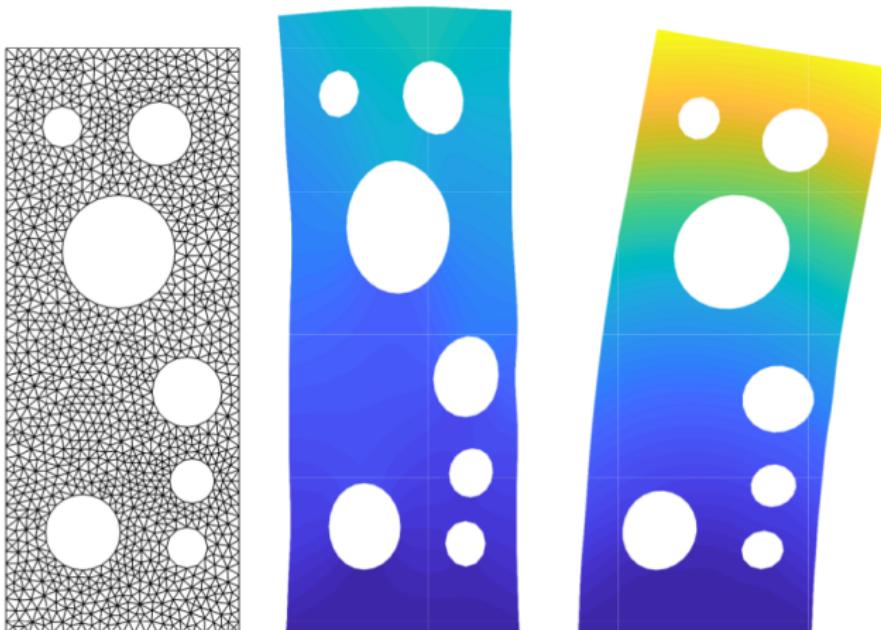
Example (3 snapshots, 40 material states, 650 mechanical states)



2D PLANE STRESS HYPER-ELASTICITY (SYNTHETIC DATA)

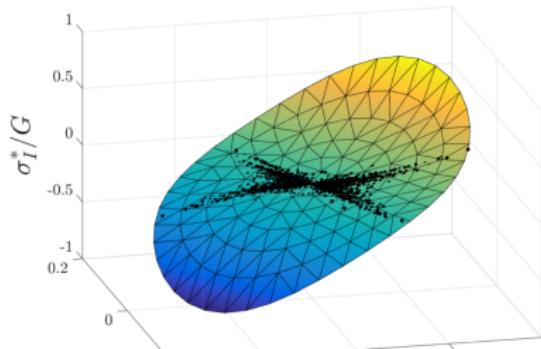
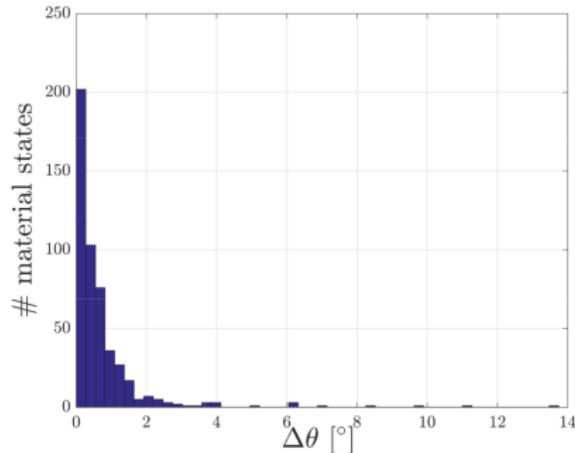
Setup (40 snapshots, 2400 elements)

$$\begin{aligned}\boldsymbol{\sigma} &= G(\boldsymbol{\epsilon} + \alpha\boldsymbol{\epsilon}^3) - p\mathbf{I}, \\ p &= -(\epsilon_{xx} + \epsilon_{yy}) - \alpha(\epsilon_{xx} + \epsilon_{yy})^3.\end{aligned}$$



2D PLANE STRESS HYPER-ELASTICITY (SYNTHETIC DATA)

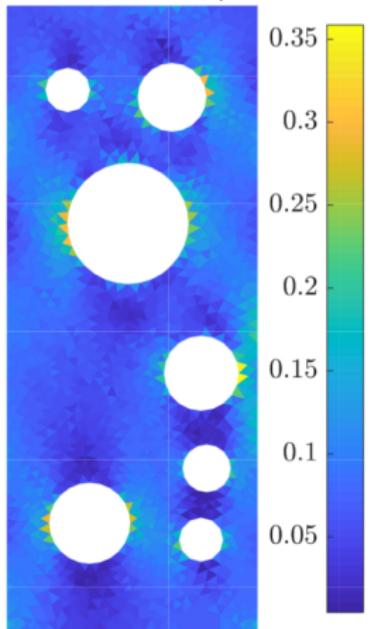
Material state database (500 states)



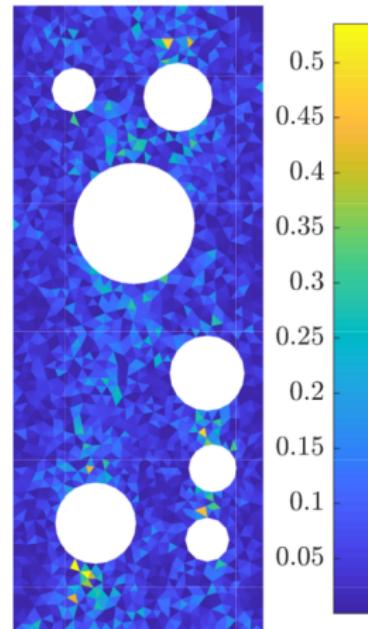
2D PLANE STRESS HYPER-ELASTICITY (SYNTHETIC DATA)

Stress reconstruction

predicted σ_{vm}/G

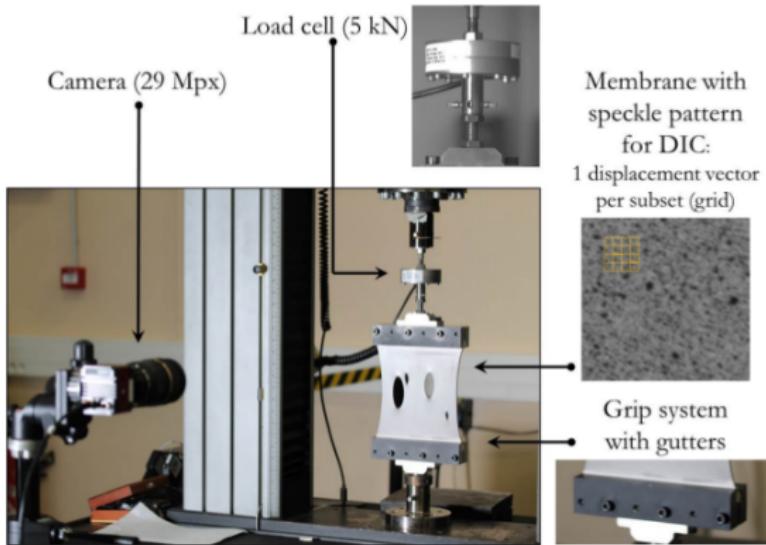


relative error in σ_{vm}



SILICONE RUBBER EXPERIMENT (MARIE DALEMAT)

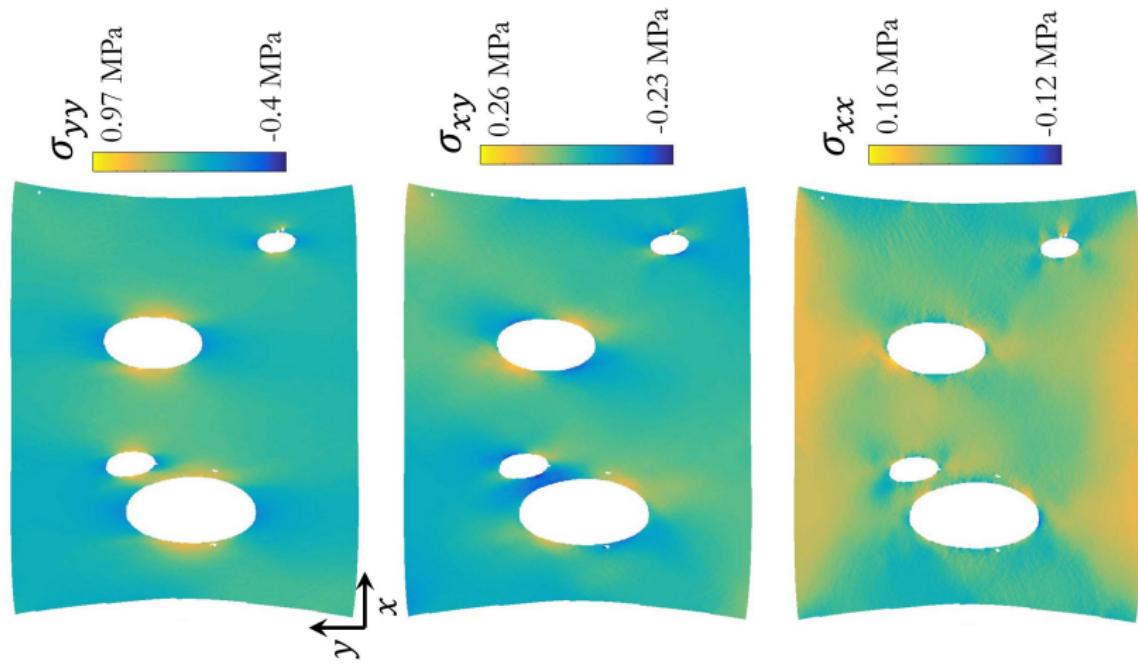
Experimental setup



- Silicone rubber RTV \approx non-linear elasticity + incompressibility
- 2D thin membrane \approx plane stress
- 150 29 MPix images

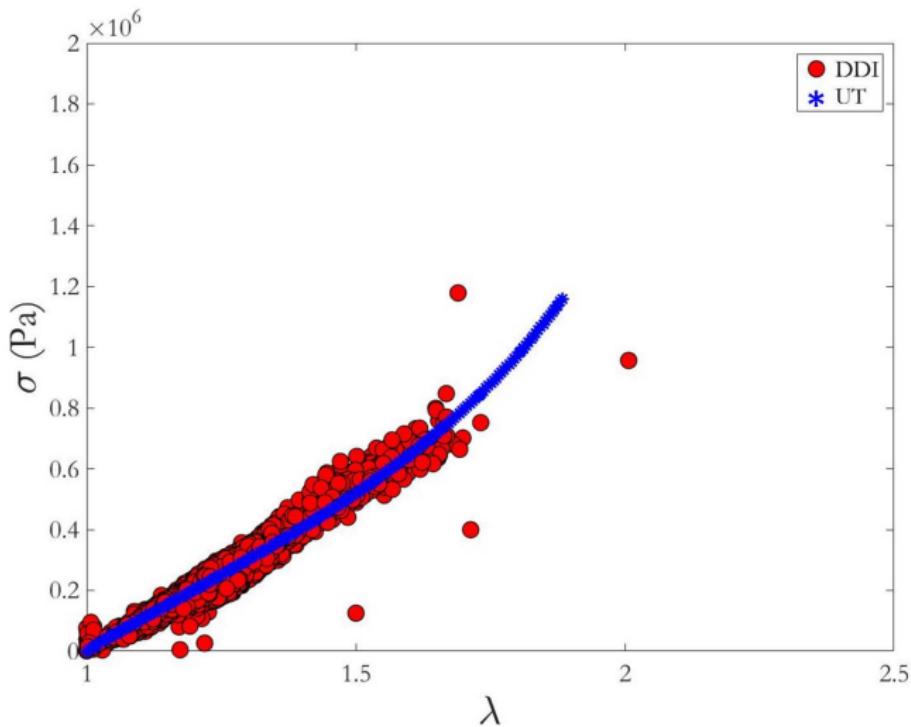
SILICONE RUBBER EXPERIMENT (MARIE DALEMAT)

Stress reconstruction (30 000 000 mechanical states)



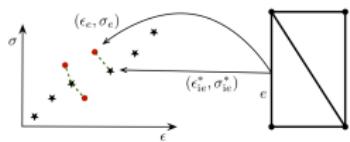
SILICONE RUBBER EXPERIMENT (MARIE DALEMAT)

Comparison with uniaxial tensile test data (10 000 material states)



First principal component

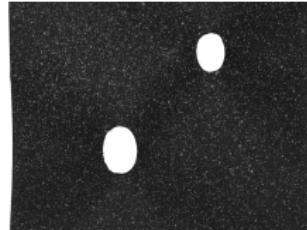
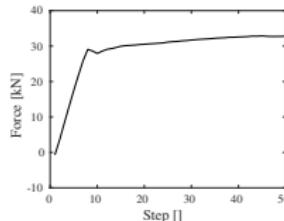
TOWARDS A NEW PARADIGM USING DATA BEYOND LAWS



- ▶ use *data* beyond *laws*
- ▶ Min distance w.r. data under BE constraint
- ▶ beyond Kirchdoerfer & Ortiz (2016)

data are obtained from *simple* experiments → limitations remain

1. Inverse data-driven → non-linear elasticity



PROBLEM STATEMENT

Assume a constitutive law has already been calibrated. One can compute the stress field $\tilde{\sigma}_e^X$ balancing the external load. It differs from the actual one

$$\sigma_e^X = \tilde{\sigma}_e^X + \bar{\sigma}_e^X$$

Only the correction $\bar{\sigma}_e^X$ is searched for.

- ▶ $\bar{\sigma}_e^X$ is self-balanced
- ▶ $\sigma_e^X = \tilde{\sigma}_e^X + L_e^j P_j^X$

The minimization problem is reformulated:

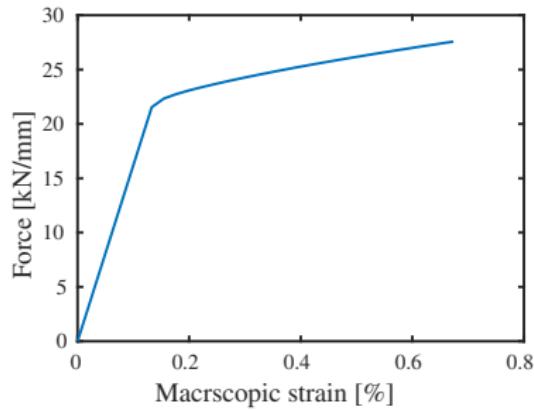
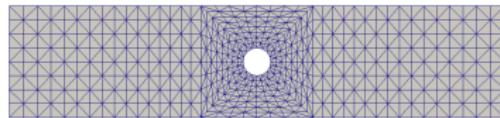
$$\text{solution} = \arg \min_{P_j^X, \epsilon_i^*, \sigma_i^*, ie^X} \mathcal{E}(\tilde{\sigma}_e^X + L_e^j P_j^X, \epsilon_i^*, \sigma_i^*, ie^X)$$

- ▶ No additional constraint
- ▶ The basis might be reduced
- ▶ The correction might be local

Specimen and model for synthetic image generation:

\mathcal{C}_0 : $E=168$ GPa, $\nu=0.25$

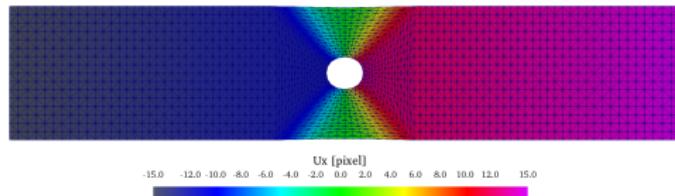
linear isotropic hardening: $\sigma_y=285$ MPa, $H=1.48$ GPa



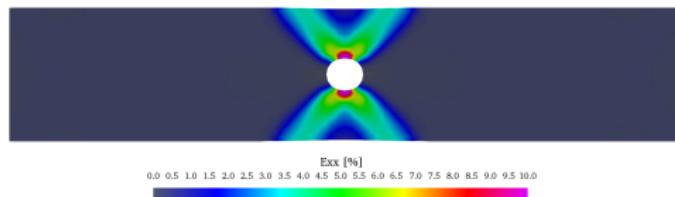
35 steps

PLASTICITY

- Displacement



- Total strain

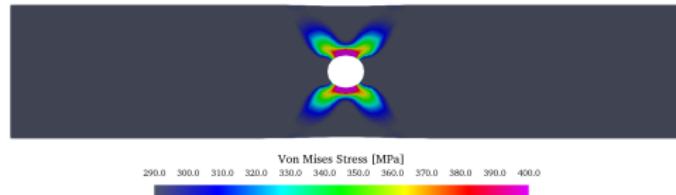


- Reduced zone (elements such that $||\epsilon|| > 0.2\%$)

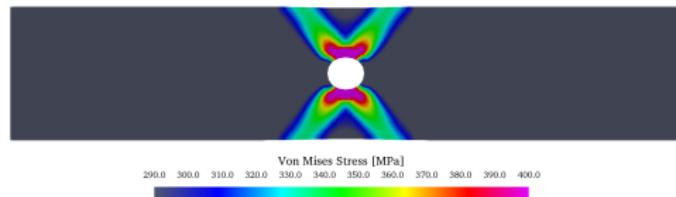


PLASTICITY

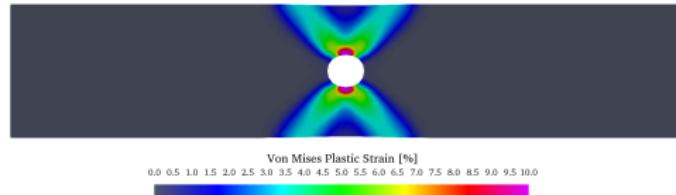
- ▶ Initial Von Mises stress (elastic)



- ▶ Corrected Von Mises stress



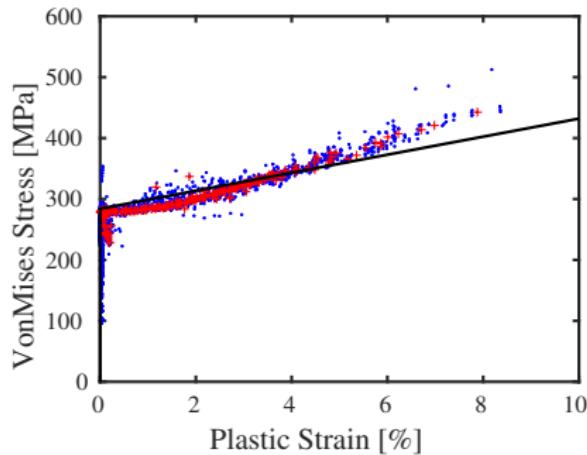
- ▶ Reconstructed Von Mises plastic strain using $\sigma_{zz} = 0$ and $Tr(\epsilon_p) = 0$



PLASTICITY

Stress/Plastic strain plot:

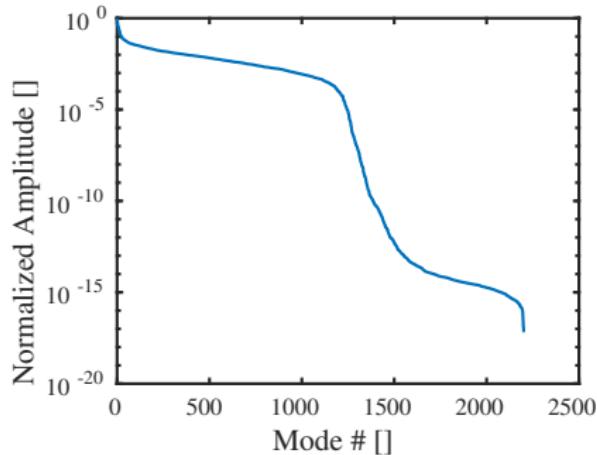
- full $\sigma, \epsilon, \epsilon_p$ history
- mechanical states, + material states



> 100000 points on a manifold of dim 6 in a space of dim 12 ($\epsilon, \sigma, \epsilon_p$)

Model reduction

$$\bar{\sigma}_e^X = L_e^j P_j^X$$



A priori $3N_e^{RD} - 2N_n^{RD}$ modes \longrightarrow only one half should be considered

CONCLUSION

Highlights

- ▶ Stress field can be reconstructed from
 - ▶ DIC displacement fields
 - ▶ force measurements
- ▶ no parametric phenomenological relation is required

Possible improvement & prospects

- ▶ best database \approx DDI for history dependent materials
- ▶ coupling with IRT (releasing the max diss. condition)
- ▶ non-proportionnal loadings
- ▶ ...

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