

# Efficient surrogate models for microstructured materials based on interaction-based material networks

Van-Dung Nguyen & Ludovic Noels

Computational & Multiscale Mechanics of Materials

<http://www.ltas-cm3.ulg.ac.be>

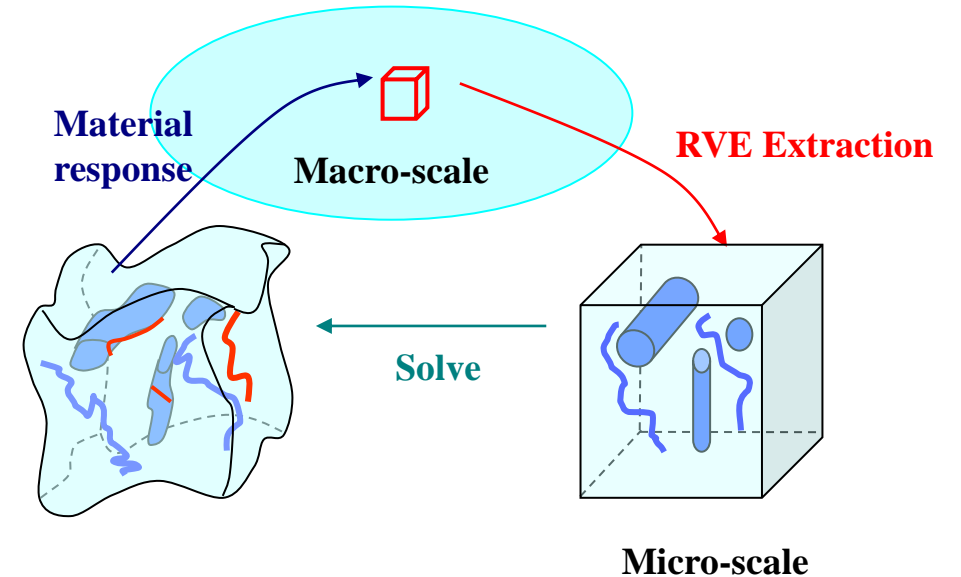
University of Liège, Belgium

[vandung.nguyen@uliege.be](mailto:vandung.nguyen@uliege.be)

[l.noels@uliege.be](mailto:l.noels@uliege.be)

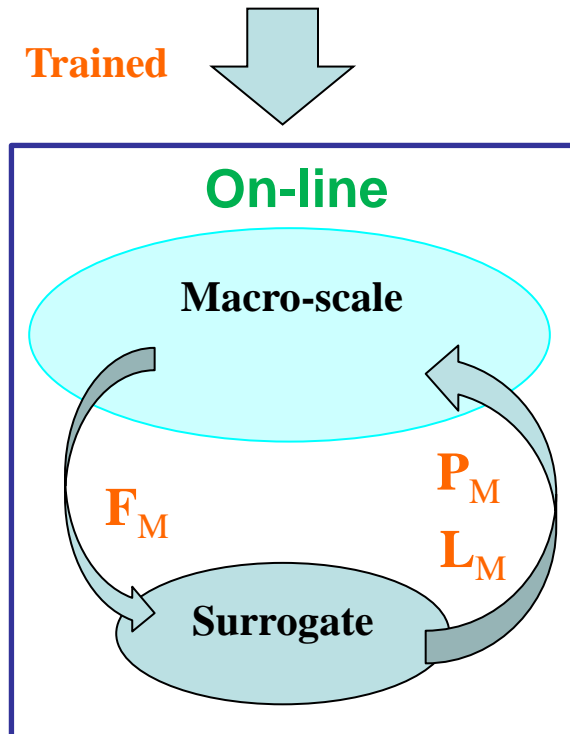
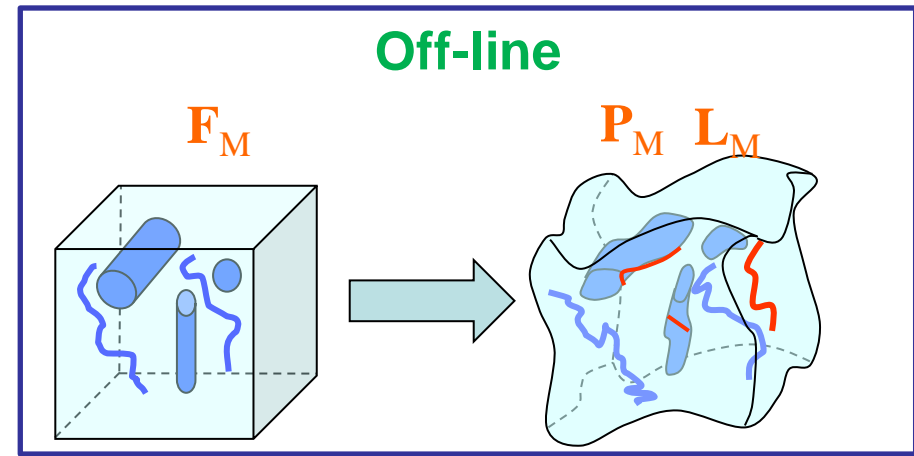
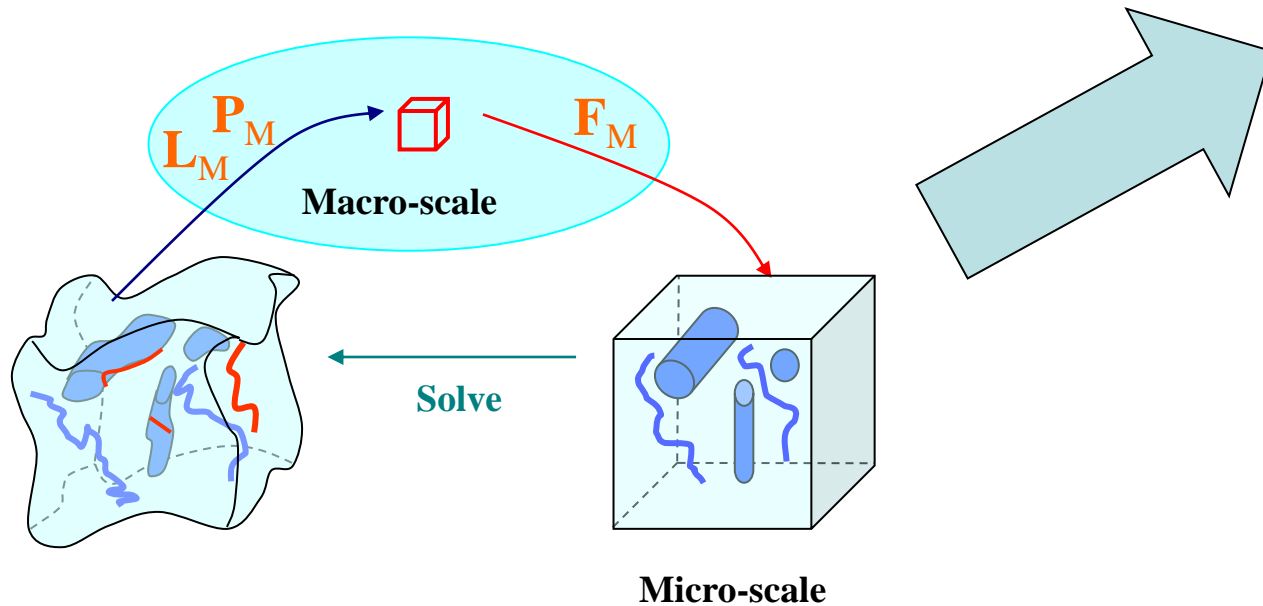
# Introduction

- **Computational homogenization (FE<sup>2</sup>)**
  - Microstructured materials
  - Two problems are solved concurrently:
    - Macro-scale: seen as a continuum
    - Micro-scale: Representative Volume Element (RVE)
- **Advantage**
  - Account directly for the micro-structural parameters (microstructure, constitutive behavior) with high accuracy.
- **Drawback**
  - Computational time & memory:
- **Solution**
  - Surrogate model of the microscopic BVP



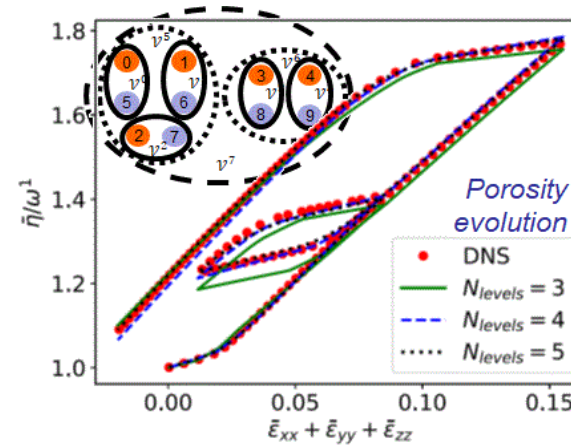
# Introduction

- **Surrogate model of the microscopic BVP**
  - Define a surrogate model
  - Off-line:
    - Construct off-line data-base (using RVE simulations)
    - Train surrogate model
  - On-line:
    - Use the trained surrogate model during analyses



# Deep material networks

- Based on thermodynamic consistency
- Possesses extrapolation capabilities in
  - Strain (history):  $\mathbf{F}_M$
  - Material parameters:  $\gamma_m$
- Emerging methodology
  - Seminal work
    - Liu, Wu, Koishi, (2019). A deep material network for multiscale topology learning and accelerated nonlinear modeling of heterogeneous materials. CMAME
  - Reformulation and use as surrogate for arbitrary material law
    - Gajek, Schneider, Böhlke, (2021). An FE–DMN method for the multiscale analysis of short fiber reinforced plastic components. CMAME
    - Nguyen V.-D., Noels, L. (2022). Interaction-based material network: A general framework for (porous) microstructured materials. CMAME

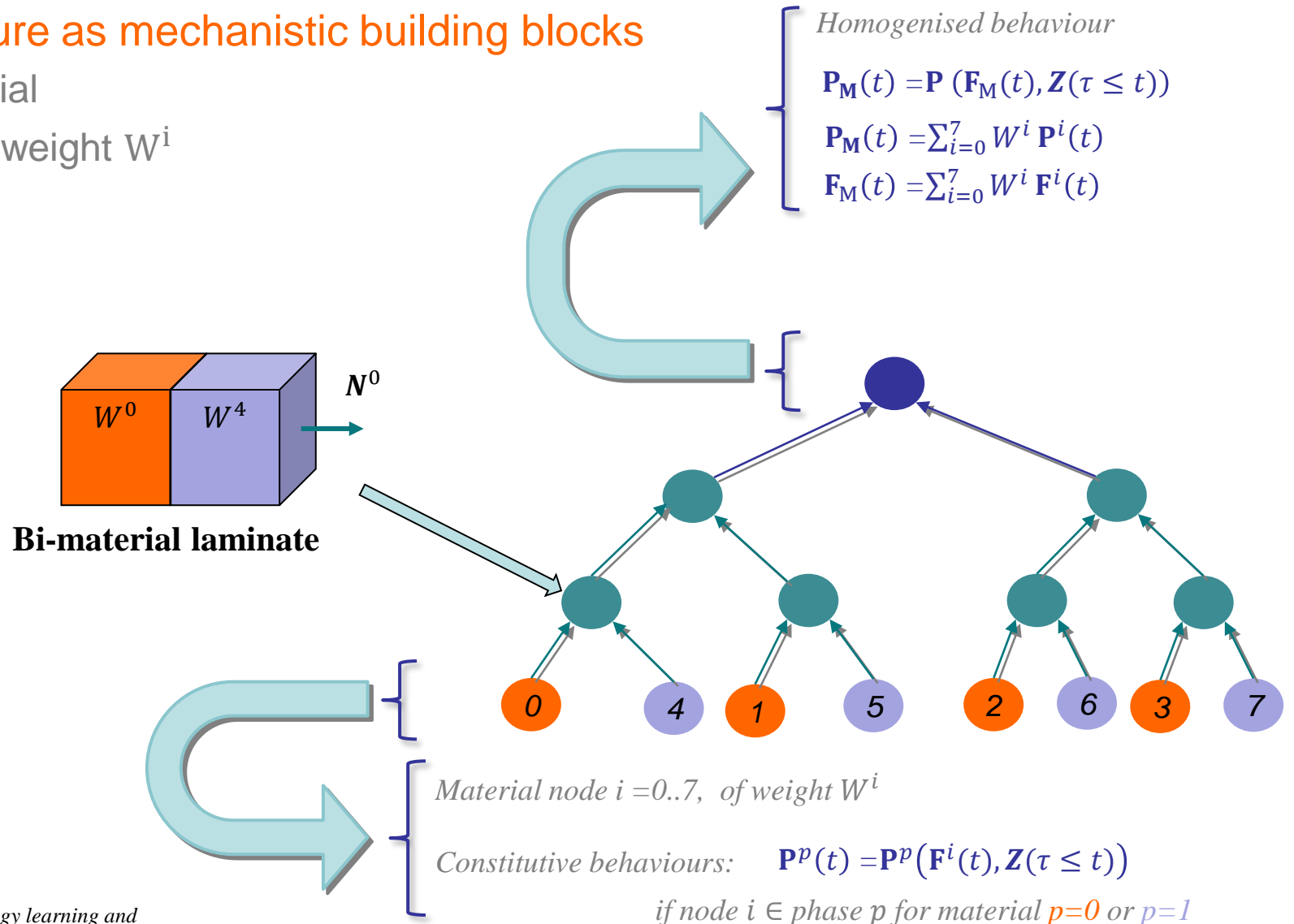


# Deep material networks

- **Bi-material laminate architecture as mechanistic building blocks**

- Example for a 2-phase material
- Material node  $i = 0, \dots, N-1$ , of weight  $W^i$

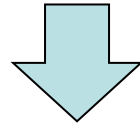
$$\sum_i W^i = 1$$



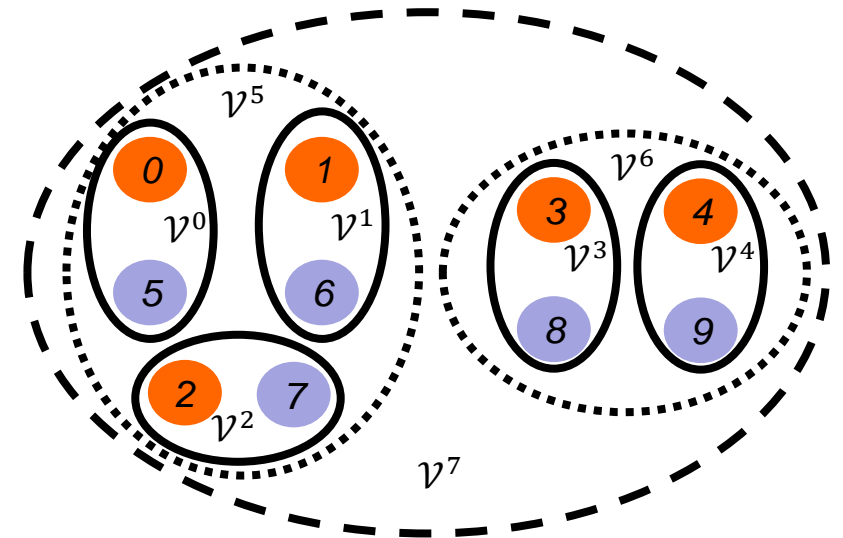
# Deep material networks

- **Our contributions**

- The solution is provided under a closed form, which is equivalent to the weak form in the finite element method
- Material networks are revisited with interactions



- Interaction-based material network:
  - A network of **interaction mechanisms**
  - General framework for (porous) microstructured materials



**Material nodes** are linked through **interactions**

- Nguyen, V. D., & Noels, L. (2022). *Micromechanics-based material networks revisited from the interaction viewpoint; robust and efficient implementation for multi-phase composites. European Journal of Mechanics - A/Solids.*
- Nguyen V.-D., Noels, L. (2022). *Interaction-based material network: A general framework for (porous) microstructured materials. CMAME*

# Interaction-based material networks – interactions & closed form solution

- Link homogenised deformation gradient to nodes

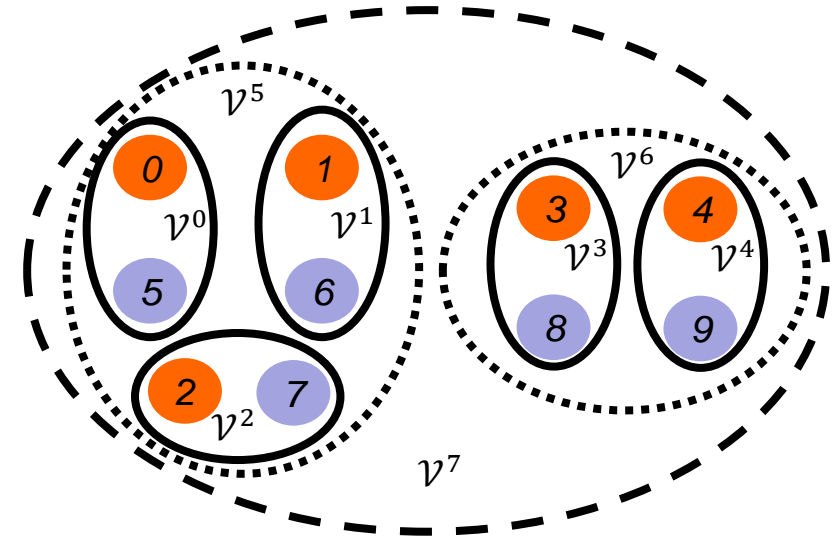
- Interaction  $k = 0..M - 1$

$$\mathbf{F}_M + \sum_{k=0}^{M-1} \alpha^{i,k} \mathbf{a}^k \otimes \mathbf{G}^k = \mathbf{F}^i, \quad i = 0..9$$

Contribution of node  $i$   
in interaction  $k$   
(parameter?)

Direction of interaction  $k$   
(parameter)

Degrees of freedom of interaction  $k$   
defining the strain fluctuation



**A material network**

- 10 material nodes
- 8 interactions

# Interaction-based material networks – interactions & closed form solution

- Link homogenised deformation gradient to nodes

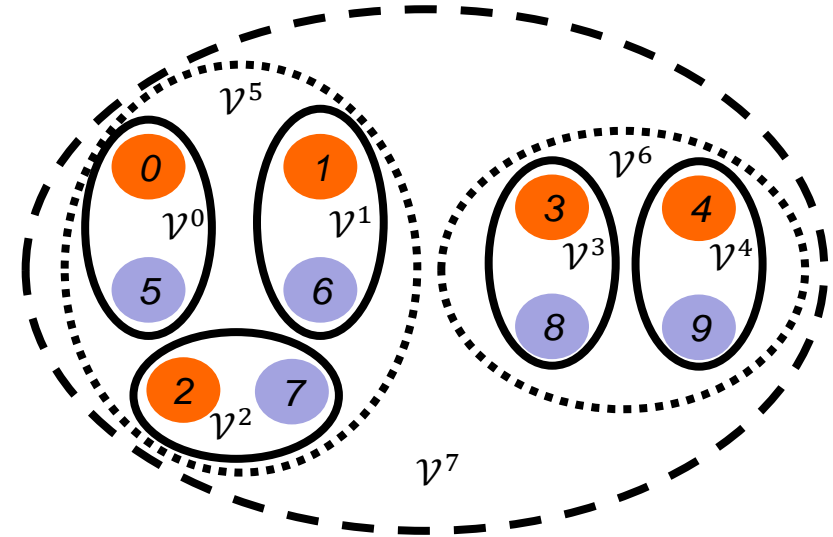
– Interaction  $k = 0..M - 1$

$$\mathbf{F}_M + \sum_{k=0}^{M-1} \alpha^{i,k} \mathbf{a}^k \otimes \mathbf{G}^k = \mathbf{F}^i, \quad i = 0..9$$

- Constraints from strain averaging

$$\mathbf{F}_M = \sum_i W^i \mathbf{F}^i \quad \Rightarrow \quad \sum_k \left( \sum_i W^i \alpha^{i,k} \right) \mathbf{a}^k \otimes \mathbf{G}^k = 0 \quad \Rightarrow \quad \sum_i W^i \alpha^{i,k} = 0$$

Constraint to be satisfied in interaction  $k$



- Weak form from Hill-Mandel

$$\mathbf{P}_M : \delta \mathbf{F}_M = \sum_i W^i \mathbf{P}^i : \delta \mathbf{F}^i \quad \Rightarrow \quad \left[ \sum_k \left( \sum_i W^i \mathbf{P}^i \alpha^{i,k} \right) \cdot \mathbf{G}^k \right] \cdot \delta \mathbf{a}^k = 0 \quad \Rightarrow \quad \left\{ \begin{array}{l} \sum_k \left( \sum_i W^i \mathbf{P}^i \alpha^{i,k} \right) \cdot \mathbf{G}^k = 0 \\ \text{Material node } i = 0..9, \text{ of weight } W^i \\ \text{Constitutive behaviours: } \mathbf{P}^p(t) = \mathbf{P}^p(\mathbf{F}^i(t), \mathbf{Z}(\tau \leq t)) \\ \text{if node } i \in \text{phase } p \text{ for material } p=0 \text{ or } p=1 \end{array} \right.$$

- Homogenized stress from stress averaging

$$\mathbf{P}_M = \sum_i W^i \mathbf{P}^i$$



# Interaction-based material network – interactions

- How to define an interaction?

- For an interaction  $k$

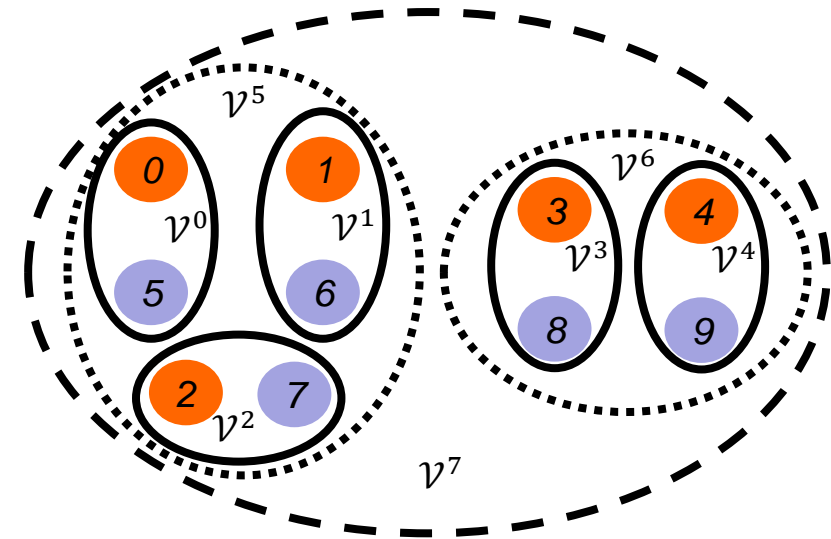
$$\sum_i W^i \alpha^{i,k} = 0$$

- Each interaction includes several nodes

- $\alpha^{i,k} = 0$  if node  $i$  does not participate the interaction  $k$

- Network architecture

- A set of material nodes
- A set of network interactions



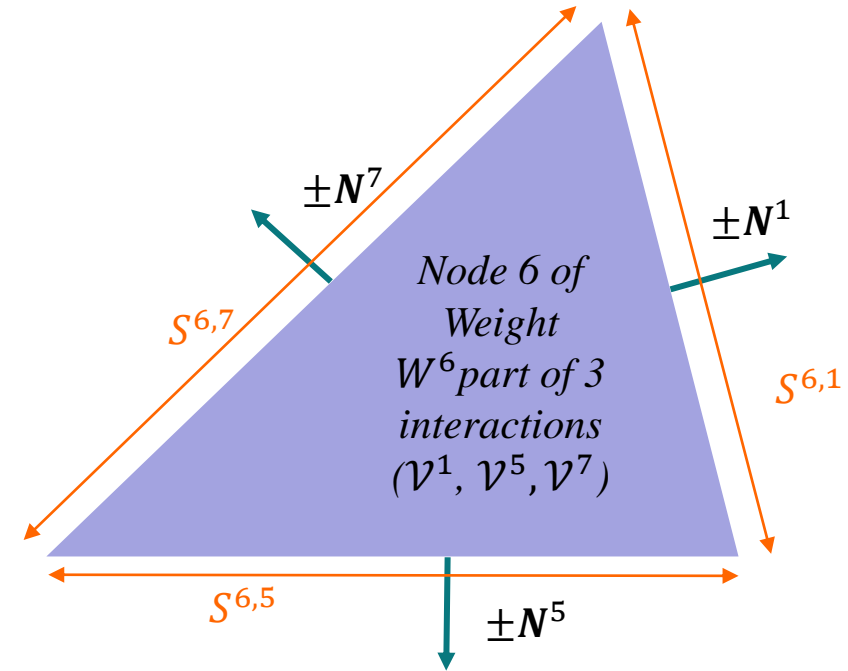
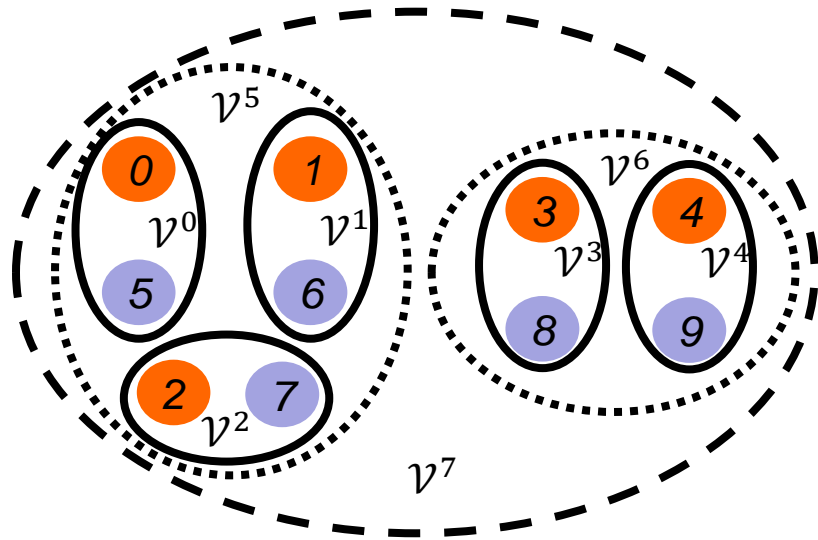
**A material network**

- 10 material nodes
- 8 interactions

# Interaction-based material networks – interactions

- Mechanistic building blocks: Polyhedra

- Interaction  $\mathcal{V}^j$ ,  $j = 0..M - 1$



- Fluctuation field:  $\mathbf{w} = \mathbf{x} - \mathbf{F}_M \cdot \mathbf{X}$

- Integration by parts on a polyhedron of volume  $V^i$  associated to node  $i$

$$\mathbf{F}_M + \frac{1}{V^i} \int_{V^i} \mathbf{w} \otimes \nabla_0 dV = \mathbf{F}^i \quad \Rightarrow \quad \mathbf{F}_M + \sum_{j:i \in \mathcal{V}^j} \frac{S^{i,j}}{V^i} \mathbf{w} \otimes (\pm \mathbf{N}^j) = \mathbf{F}^i$$

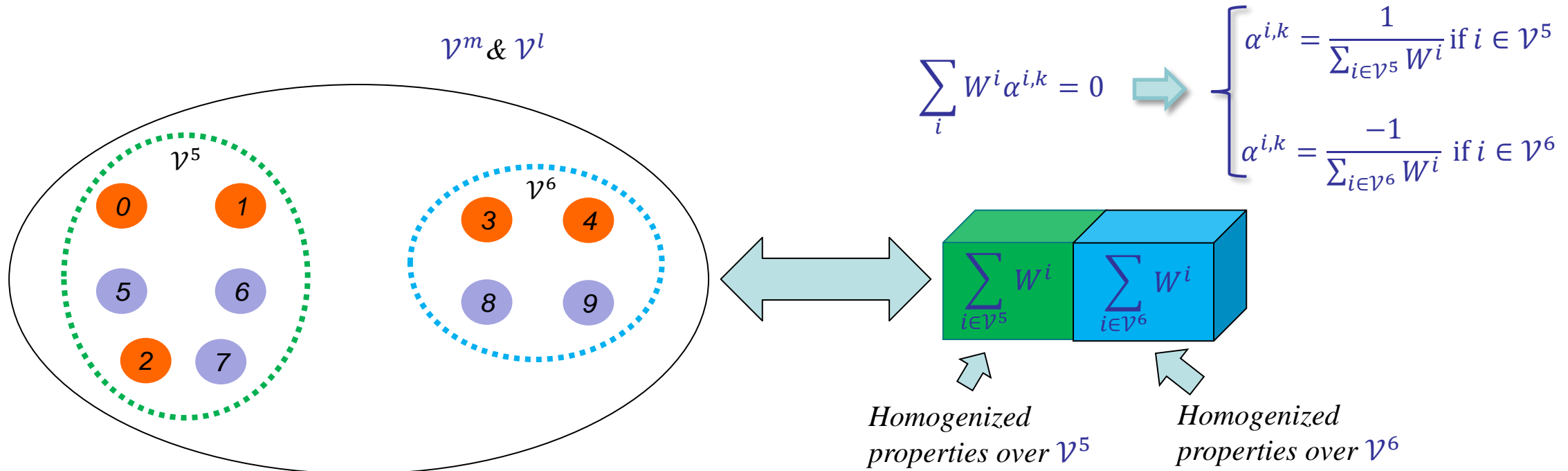
- To be compared with the interactions

$$\mathbf{F}_M + \sum_{k=0}^{M-1} \alpha^{i,k} \mathbf{a}^k \otimes \mathbf{G}^k = \mathbf{F}^i$$

- $\alpha^{i,k}$  is the weighted surface of a polyhedron face (parameter to be identified)
- $\mathbf{G}^k$  is the inward or outward normal of the polyhedron face (parameter to be identified)
- $\mathbf{a}^k$  is the fluctuation field (degree of freedom for online simulations)

# Interaction-based material network – interactions

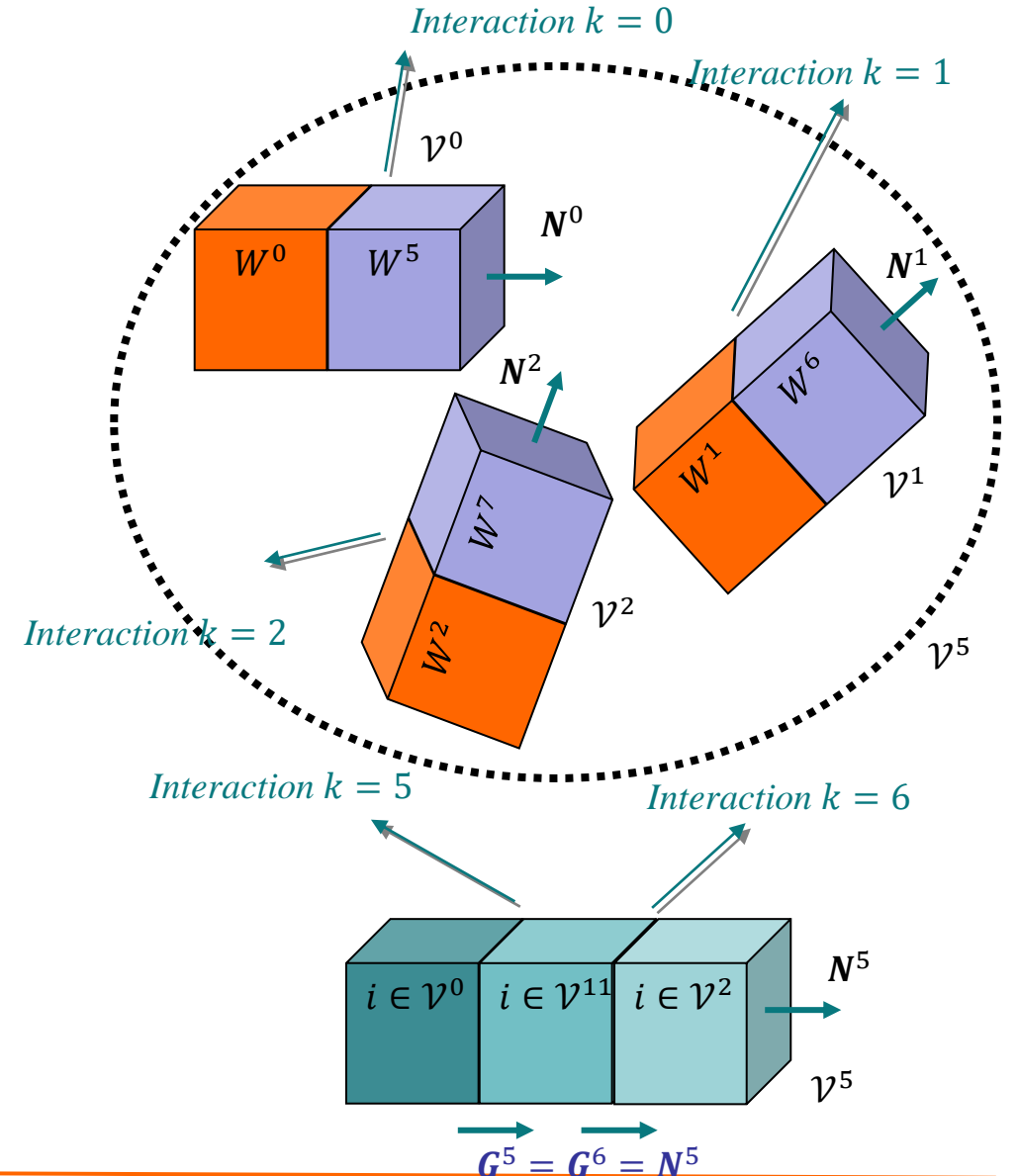
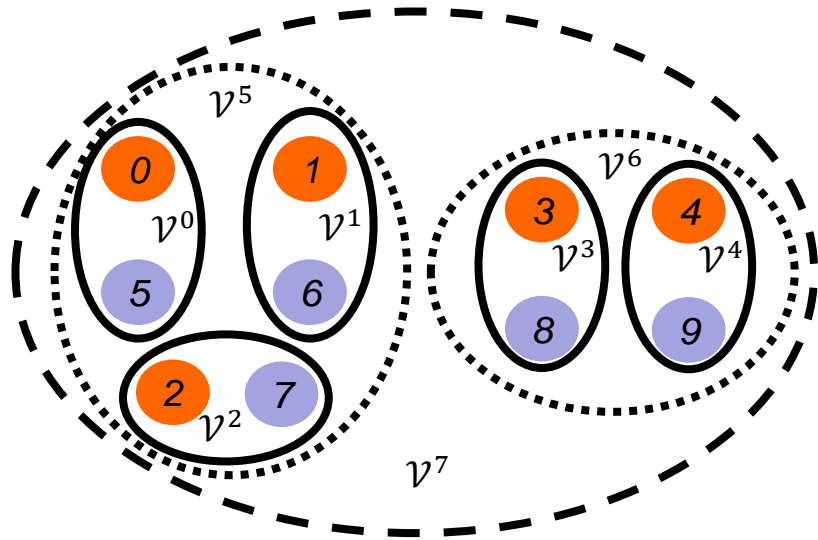
- Mechanistic building blocks: we can recover laminate building blocks
  - Between two groups of nodes



- Nguyen, V. D., & Noels, L. (2022). Micromechanics-based material networks revisited from the interaction viewpoint; robust and efficient implementation for multi-phase composites. *European Journal of Mechanics - A/Solids*.

# Interaction-based material network – interactions

- Mechanistic building blocks: Laminate



Trainable parameters:

- Weights:  $W^i, i = 0..9$
- Direction for an interaction  $G^k \Rightarrow N^j, j = 0..7$

# Interaction-based material network – interactions

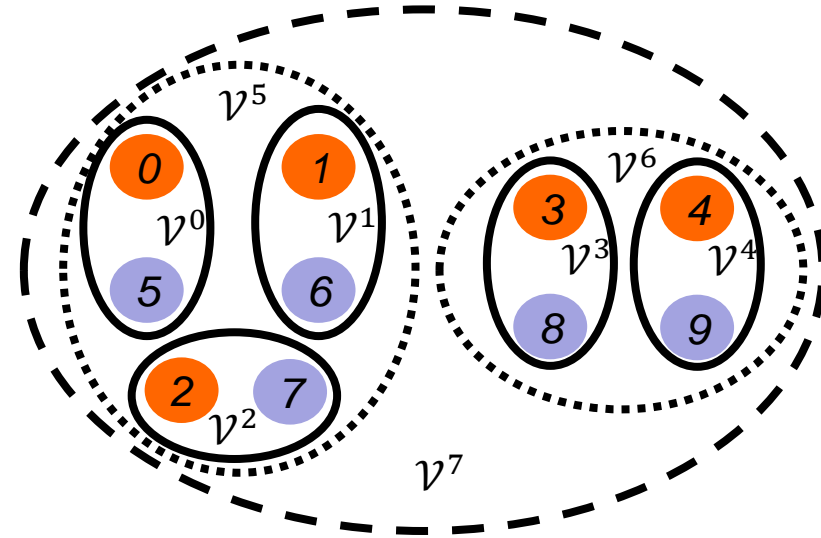
- Mechanistic building blocks: Full interaction

- Interaction  $\mathcal{V}^j$  as a full interaction

- Mechanism  $j$  is a full interaction if satisfying

$$\sum_i W^i \alpha^{i,k} = 0$$

- $\alpha^{i,k} \forall i, k$  are considered as fitting parameters



- More trainable parameters than laminate-based interactions

*Trainable parameters*

$$\mathcal{L} = [W_i, \alpha_{ij}, \mathbf{G}_j \text{ with } i = 0, \dots, N - 1, j = 0, \dots, M - 1]$$

satisfying 
$$\sum_{i=0}^{N-1} W_i \alpha_{ij} = 0 \forall j$$

} Represent RVE geometry

# Interaction-based material network – offline training

- **Linear elastic training**

- The tangent at zero strain is considered:

$$\mathbf{L}_M = \frac{\partial \mathbf{P}_M}{\partial \mathbf{F}_M} \text{ at } \mathbf{F}_M = \mathbf{I}$$

$$\mathbf{L}_M = \mathbf{L}_M(\mathbf{L}_0, \dots, \mathbf{L}_{P-1}; \mathcal{L})$$

*Elastic tangent tensors of  $P$  underlying phases*

*Trainable parameters*

- Offline data
  - RVE & microscopic boundary condition
  - Elastic tangent tensors of  $P$  underlying phases randomly generated
  - Homogenized tangent obtained by computational micromechanics
- A loss function is defined
- Gradient-descent optimizer to minimize this loss function

- **Nonlinear training**

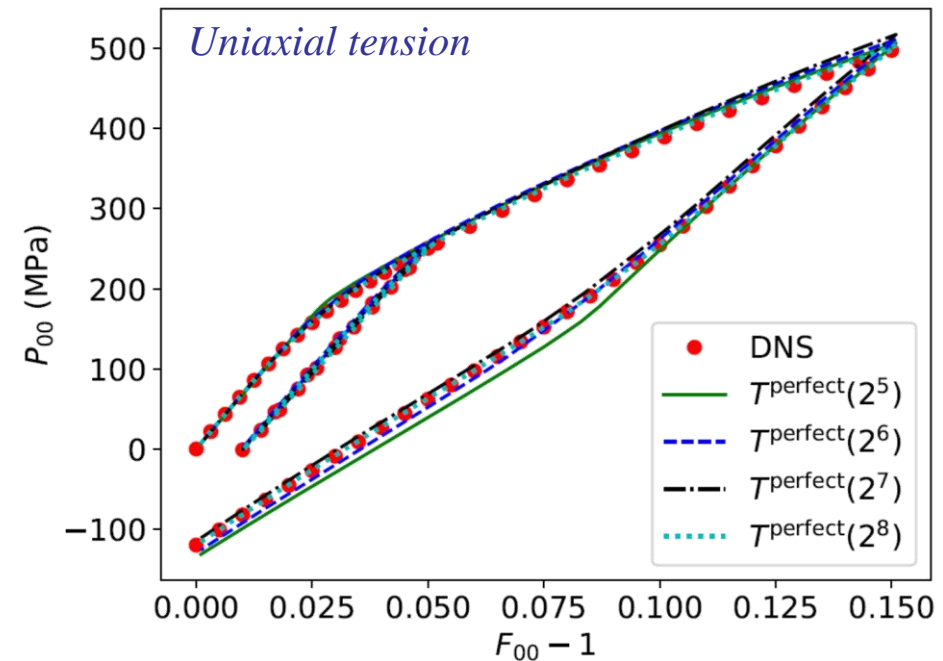
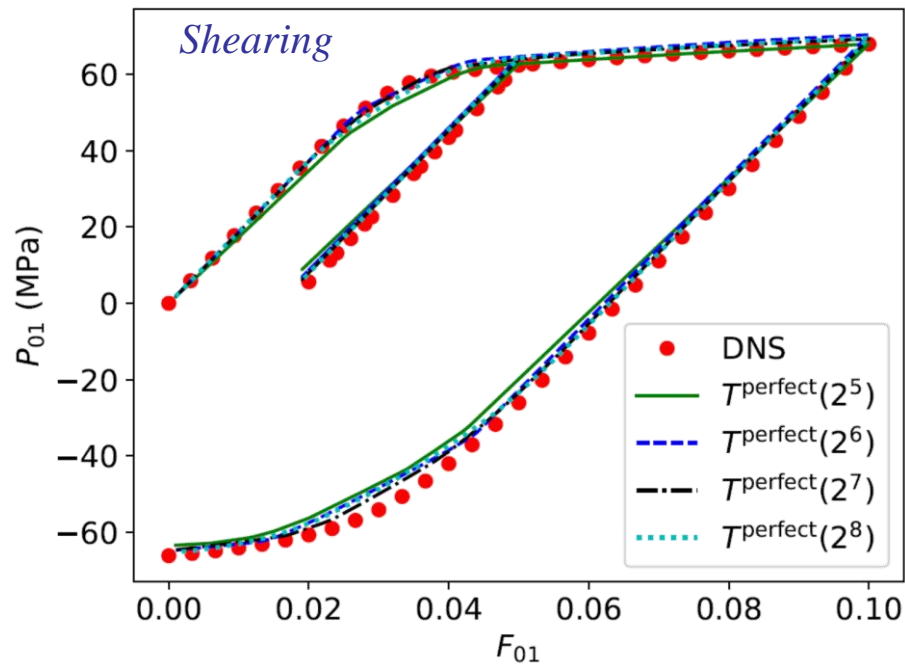
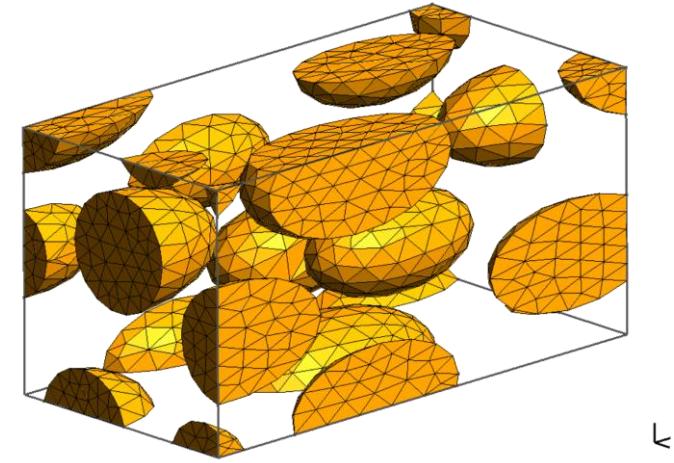
- Consider history path dependent

$$\mathbf{P}_M(t) = \mathbf{P}_M(\mathbf{F}_M(\tau), \tau \leq t; \mathcal{L})$$

- Offline data
  - RVE & microscopic boundary condition
  - Inputs : strain paths  $\mathbf{F}_M(\tau), \tau \leq t$
  - Output: stress path  $\mathbf{P}_M(\tau), \tau \leq t$  obtained by computational micromechanics
- A loss function is defined
- Gradient-descent optimizer to minimize this loss function

# Interaction-based material network as surrogate model

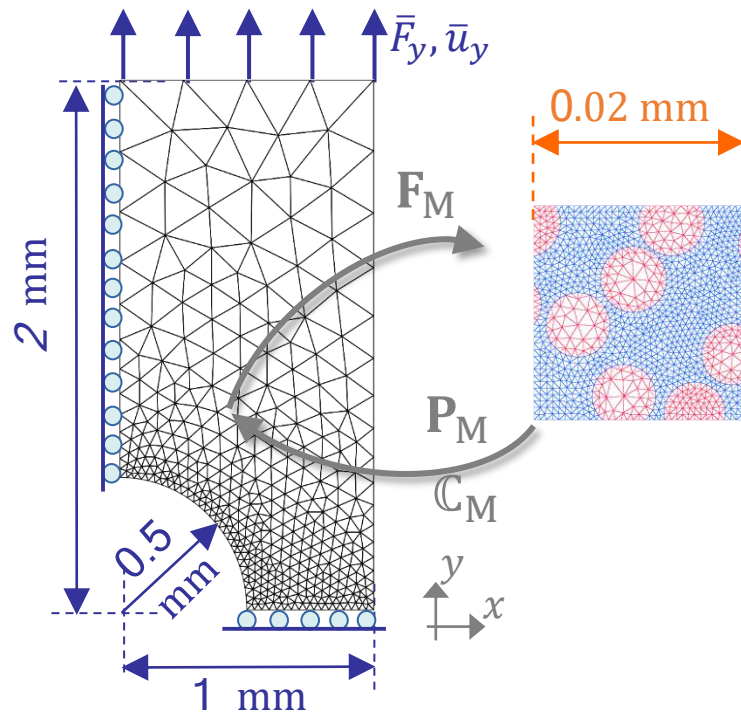
- Online stage on a particle-reinforced composite
  - Properties
    - Elastic inclusions & elasto-plastic matrix
  - Laminate-based interactions
  - Linear elastic training



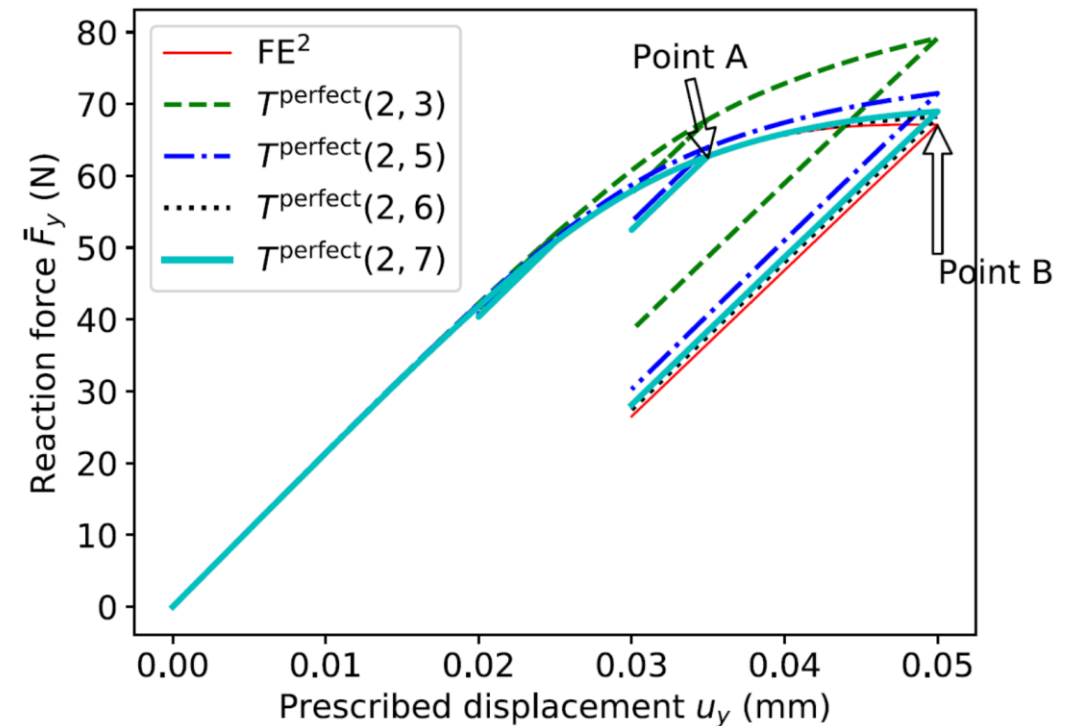
# Interaction-based material network as surrogate model

- Multiscale simulation

- Elasto-plastic composite RVE
- Comparison FE<sup>2</sup> vs. Material network-surrogate
- Laminate-based interactions
- Linear elastic training



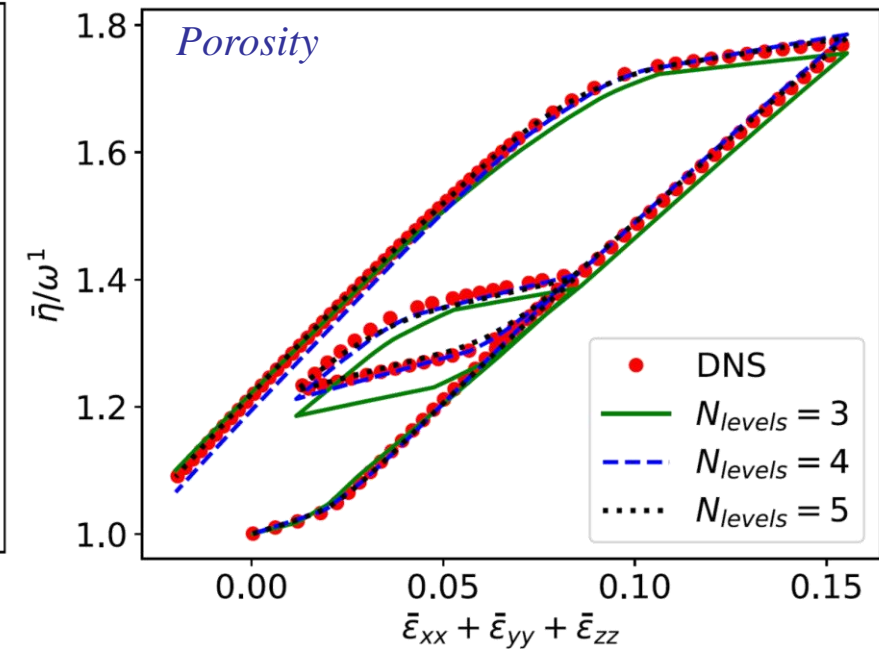
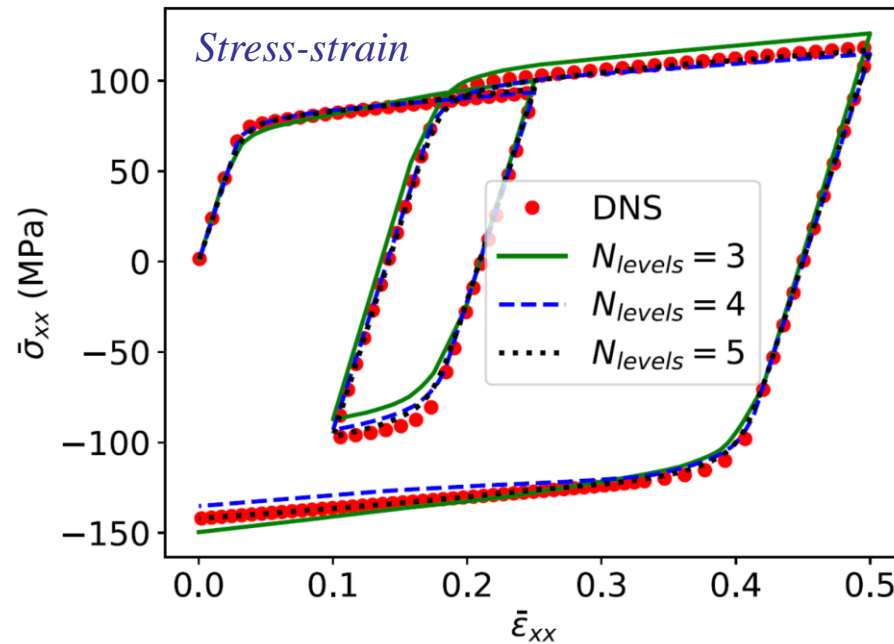
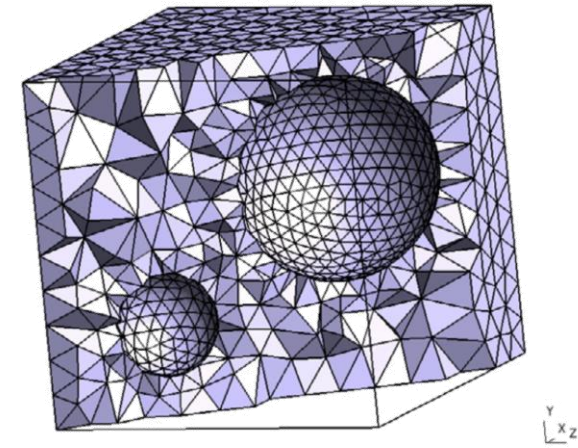
Off-line	FE <sup>2</sup>	FE-DMN
Data generation	-	10 min.-cpu
Training	-	2 min.-cpu
On-line	FE <sup>2</sup>	FE-DMN
Simulation	18000 h-cpu	½ to 34 h-cpu





# Interaction-based material network as surrogate model

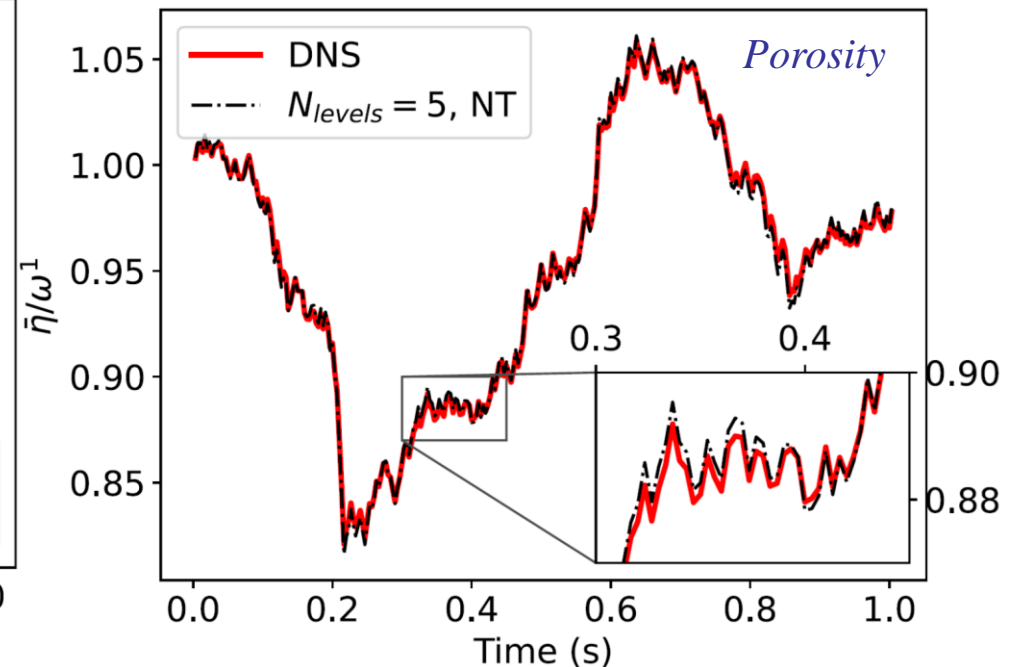
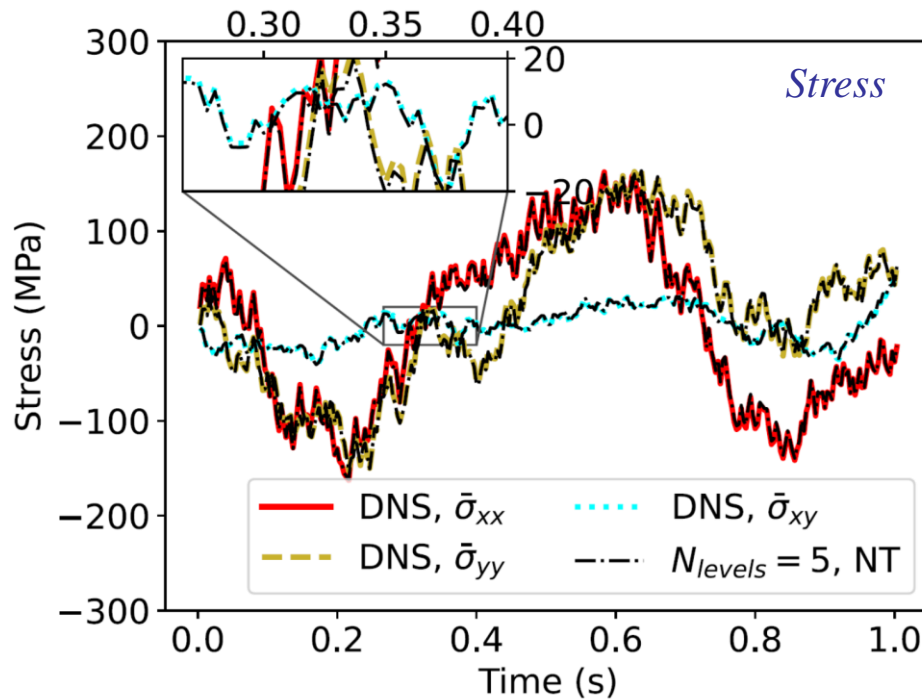
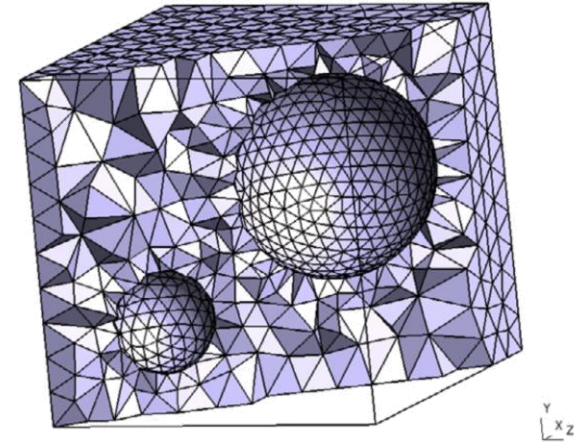
- Online stage on a porous material
  - Properties
    - Elasto-plastic matrix
    - Small strain
  - Full interactions
  - Non-linear training



# Interaction-based material network as surrogate model

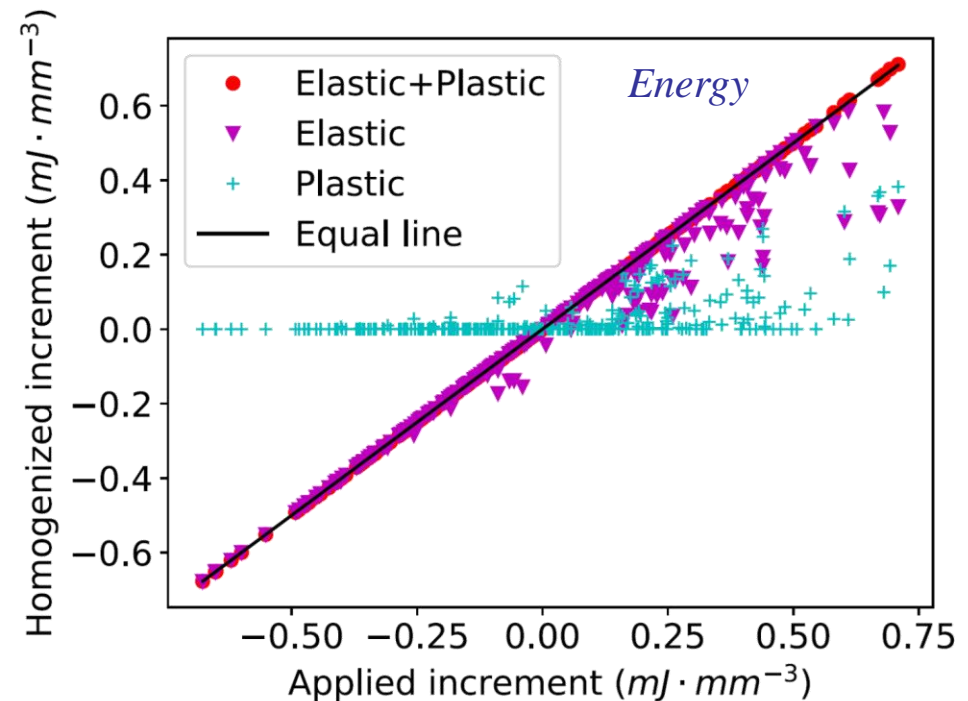
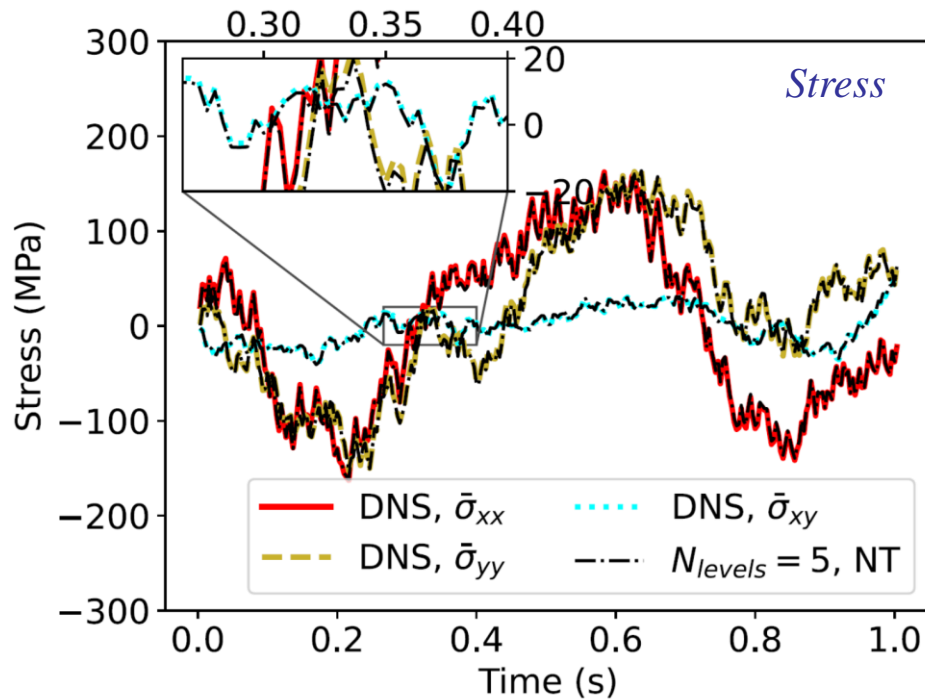
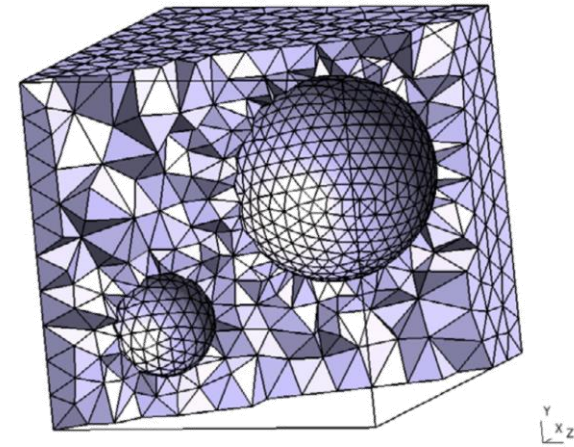
- Online stage on a porous material

- Properties
  - Elasto-plastic matrix
  - Small strain
- Full interactions as mechanistic building blocks
- Non-linear training with Material 1, on-line material 2
- Random loading



# Interaction-based material network as surrogate model

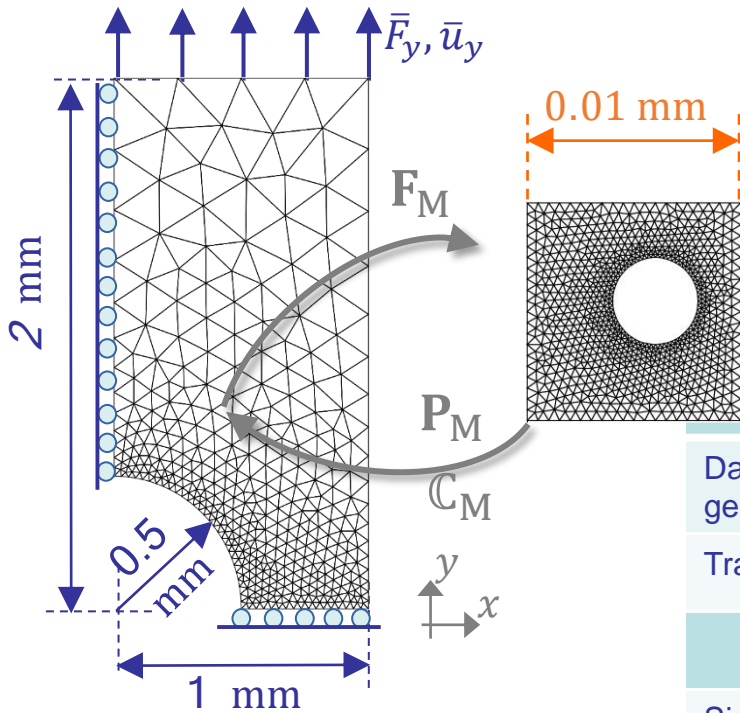
- Online stage on a porous material
  - Properties
    - Elasto-plastic matrix
    - Small strain
  - Full interactions as mechanistic building blocks
  - Non-linear training
  - Thermodynamically consistent



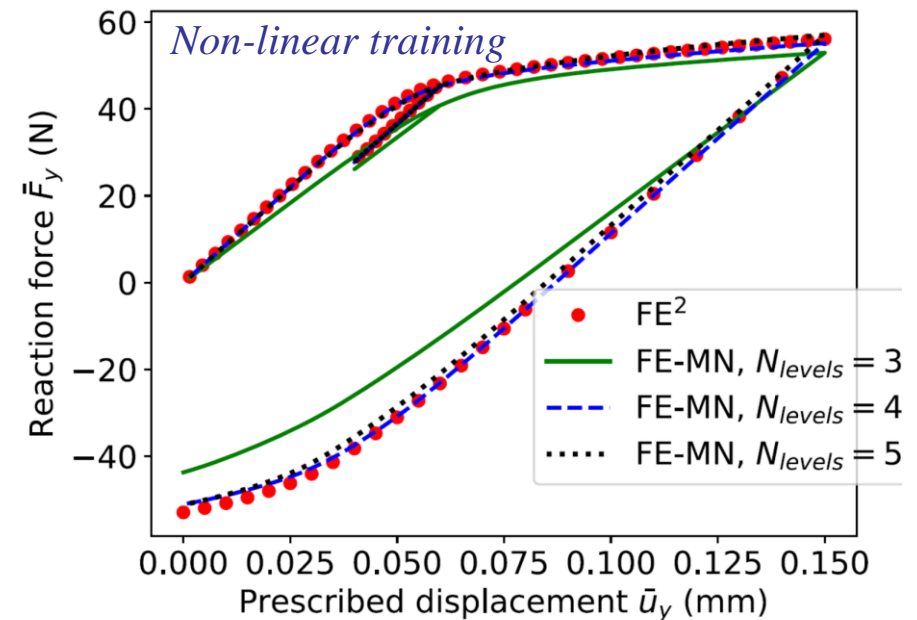
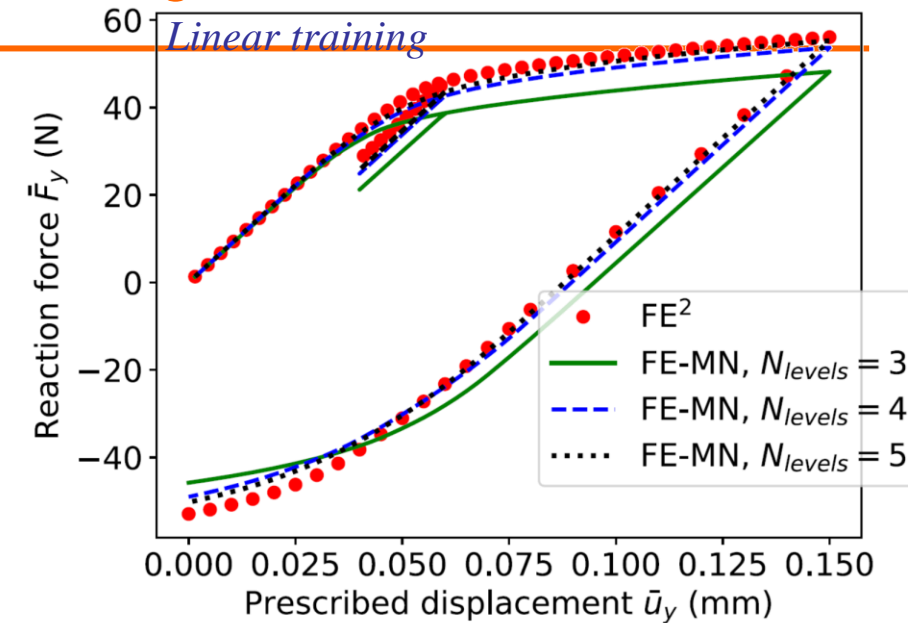
# Interaction-based material network as surrogate model

- Multiscale simulation

- Comparison FE<sup>2</sup> vs. Material network-surrogate
- Full interactions as mechanistic building blocks



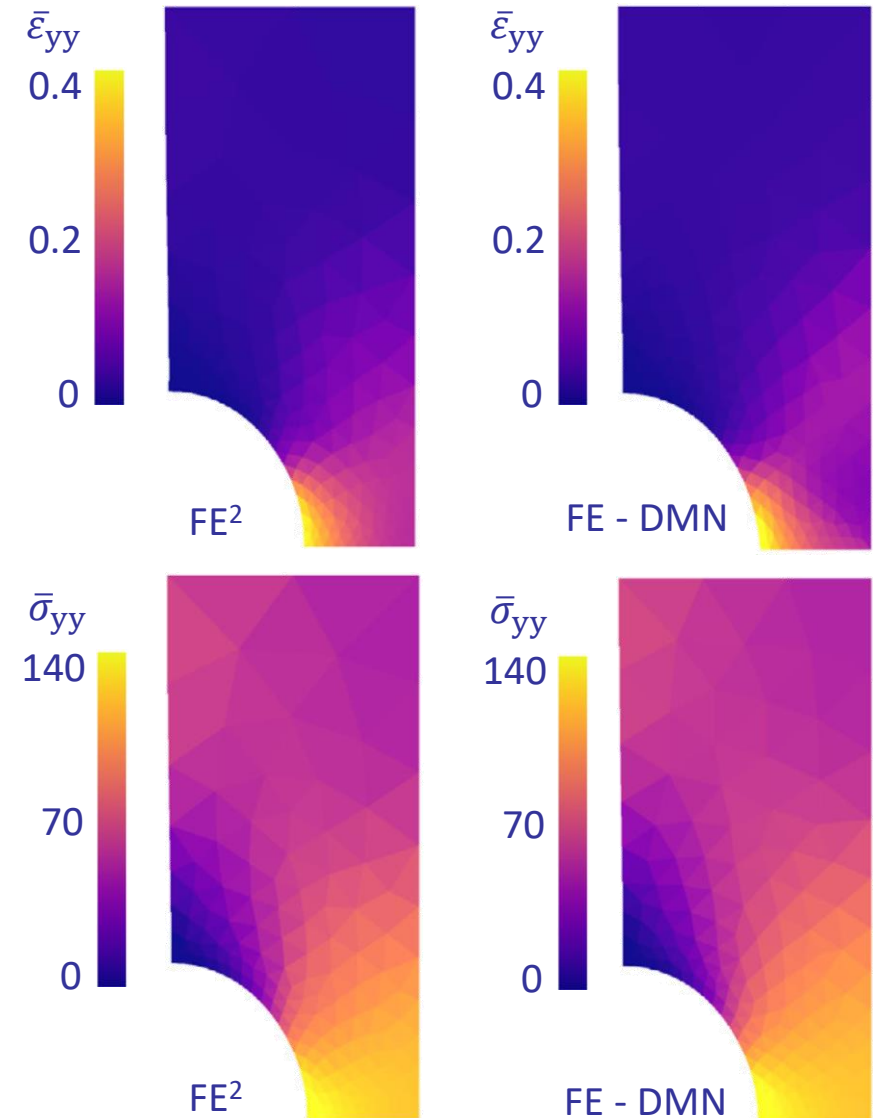
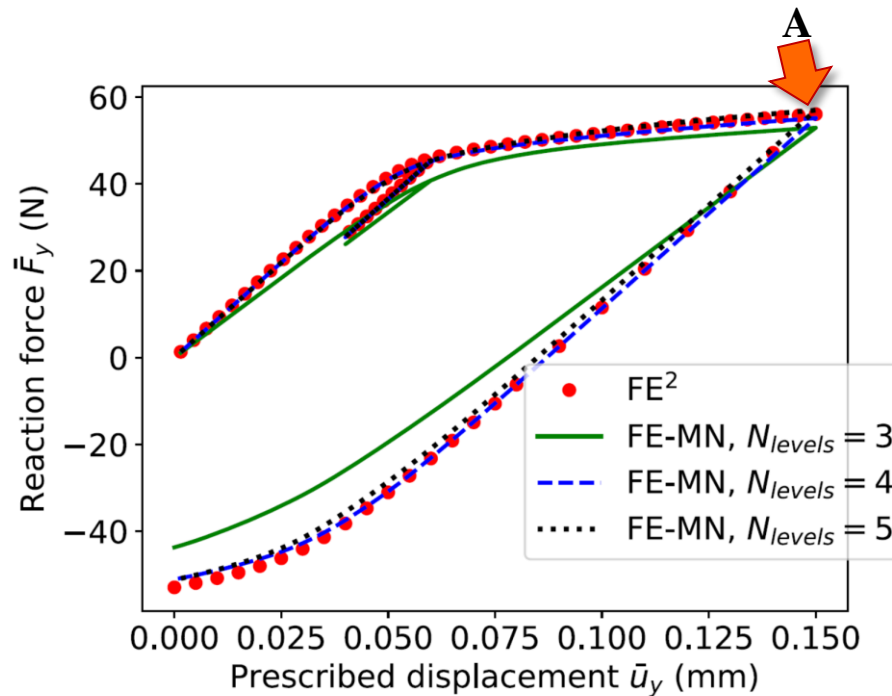
	Off-line	FE <sup>2</sup>	FE-DMN
Data generation	-	-	0.04 (linear) – 3.5 (non-linear) hour.-cpu
Training	-	-	0.16-20 hours.-cpu
On-line	FE <sup>2</sup>	FE-DMN	
Simulation	7200 h-cpu	0.1 to 1 h-cpu	



# Interaction-based material network as surrogate model

- Multiscale simulation

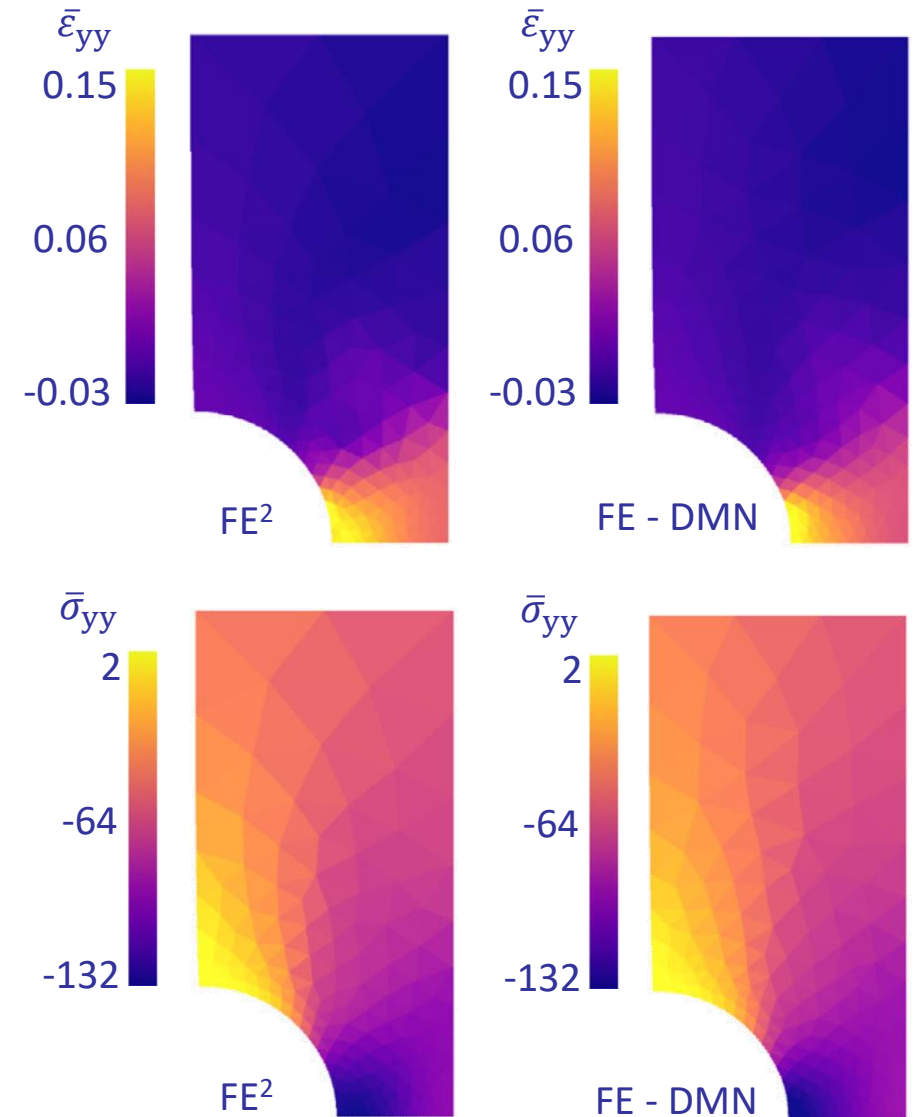
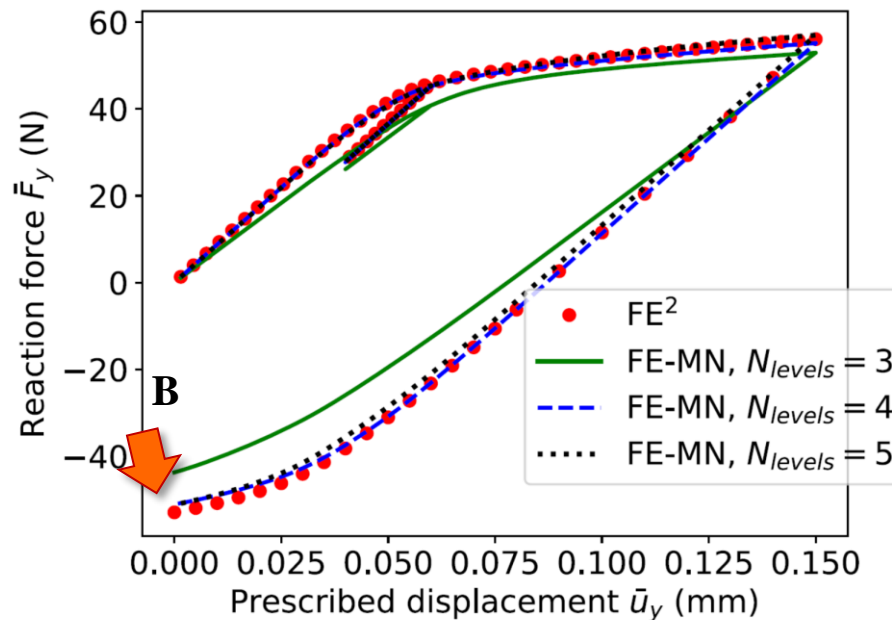
- Stress-strain distribution at point **A**
- For  $2^5$  material nodes
- Full interactions as mechanistic building blocks
- Non-linear training



# Interaction-based material network as surrogate model

- Multiscale simulation

- Stress-strain distribution at point **B**
- For  $2^5$  material nodes
- Full interactions as mechanistic building blocks
- Non-linear training



# Conclusions and perspectives

---

- **Interaction-based material network**
  - a general framework to build surrogate models for micro-structured materials
  - Satisfy all requirements of a truly microscopic boundary value problem including the stress and strain averaging principles and the Hill–Mandel energetically consistent condition
  - Efficient training procedures
  - Trained material networks with the ones of the direct numerical simulations in both contexts of virtual testing and multiscale simulations.

---

## Thank you for your attention

- *Nguyen, V. D., & Noels, L. (2022). Micromechanics-based material networks revisited from the interaction viewpoint; robust and efficient implementation for multi-phase composites. European Journal of Mechanics - A/Solids.*
- *Nguyen V.-D., Noels, L. (2022). Interaction-based material network: A general framework for (porous) microstructured materials. CMAME*
- *Data of " V. D. Nguyen and L. Noels. "Interaction-based material network: a general framework for (porous) microstructured materials." Computer Methods in Applied Mechanics and Engineering", <https://doi.org/10.5281/zenodo.5568832>*
- *Data of " V. D. Nguyen and L. Noels. "Micromechanics-based material networks revisited from the interaction viewpoint; robust and efficient implementation for multi-phase composites." European Journal of Mechanics. A, Solids 91 (January 2022): 104384. ", <https://doi.org/10.5281/zenodo.4743654>*