



domminio

Digital method for improved manufacturing of next-generation multifunctional airframe parts

Multiscale Modeling of Composites: Towards AI assisted virtual testing of composites

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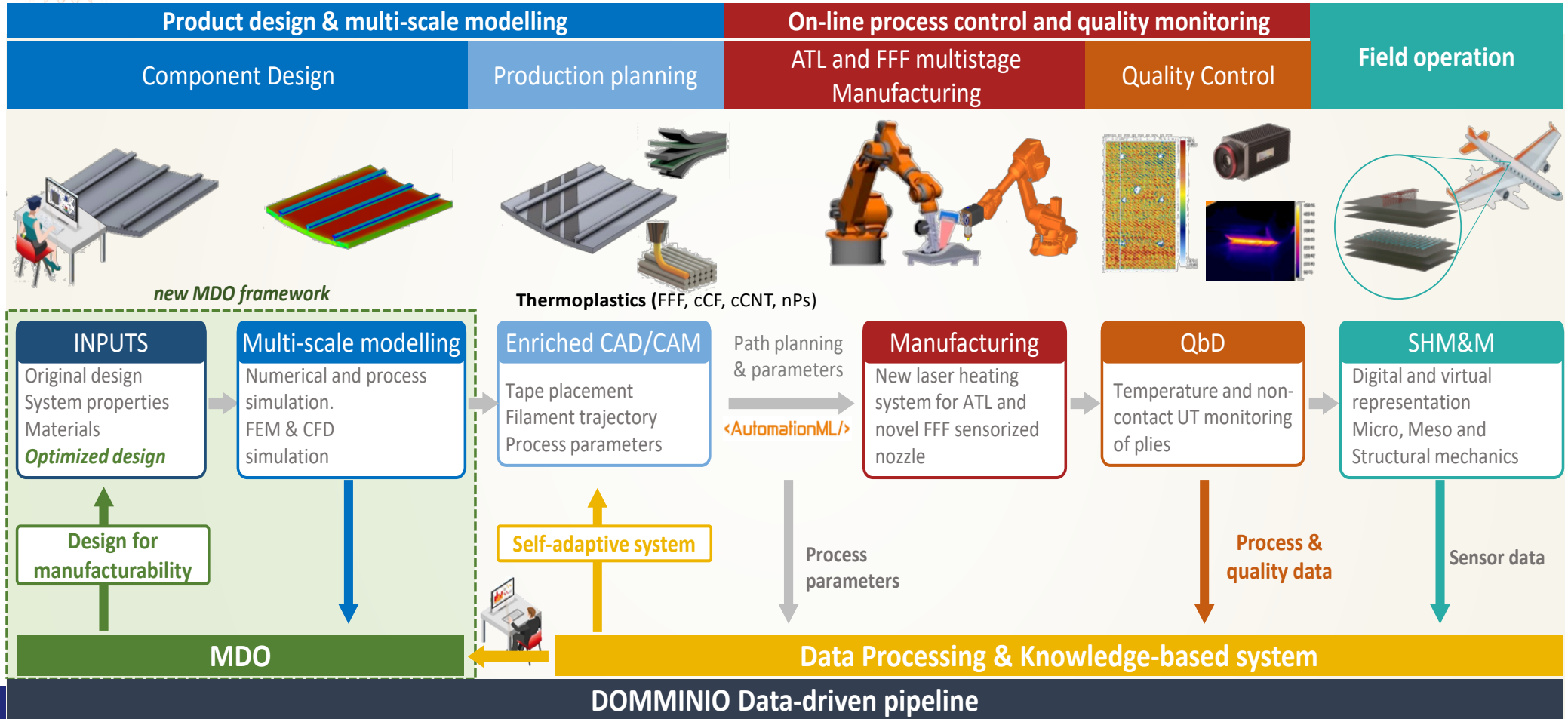
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CM3

Barcelona, Spain, 22-24 November 2021





VIRTUAL TESTING

- Physically-based ✓
- Time consuming ✗
- On-the-fly ✗

high

Use of scientific knowledge

continuum
solid & fluid
mechanics, chemistry

Hybrid methods
Simulation guided
data-science

deep learning & big data

SIMULATION GUIDED AI

- Trustworthy AI (physically-constrained)
- Reduce of tests for data-sets

ARTIFICIAL INTELLIGENCE

- Hidden patterns ✓
- Black-box ✗
- Experimental trials are expensive ✗

low

Use of scientific data

high



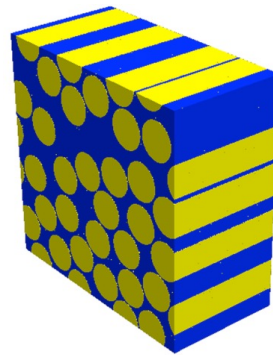
🕒 Fiber-reinforced composites present a simple -but efficient- hierarchical structure that leads to tough materials from brittle constituents.

dominant
length scale

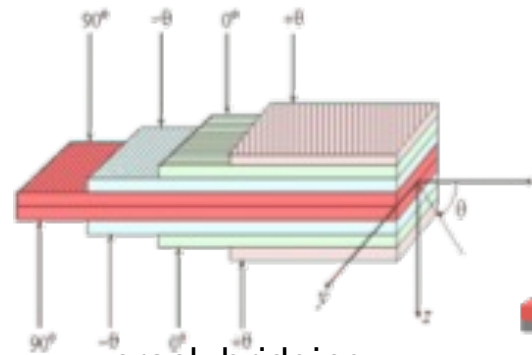
fiber diameter
 $\approx 10 \mu\text{m}$

ply thickness
 $150\text{-}300 \mu\text{m}$

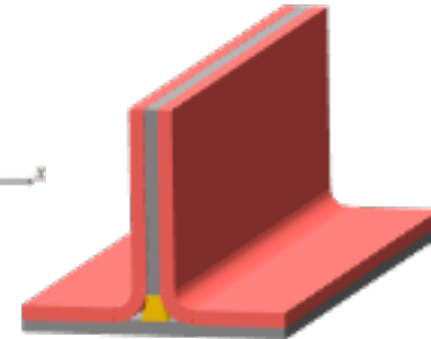
laminate thickness
 $2\text{-}20 \text{ mm}$



matrix shear yielding
interface decohesion



crack bridging
fiber pull-out
fiber kinking
interply decohesion

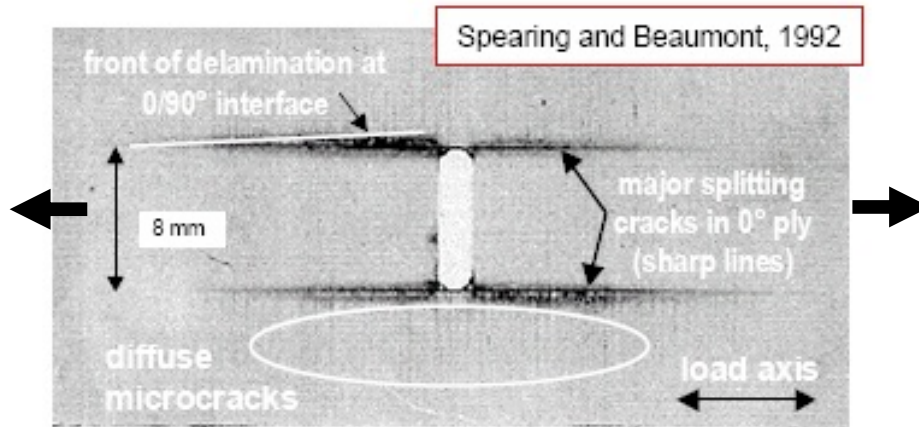
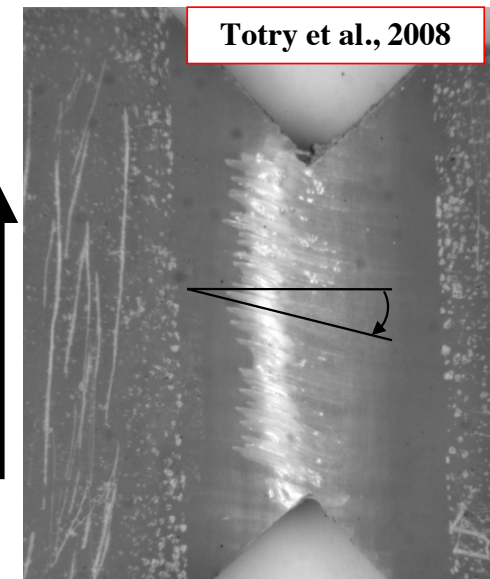
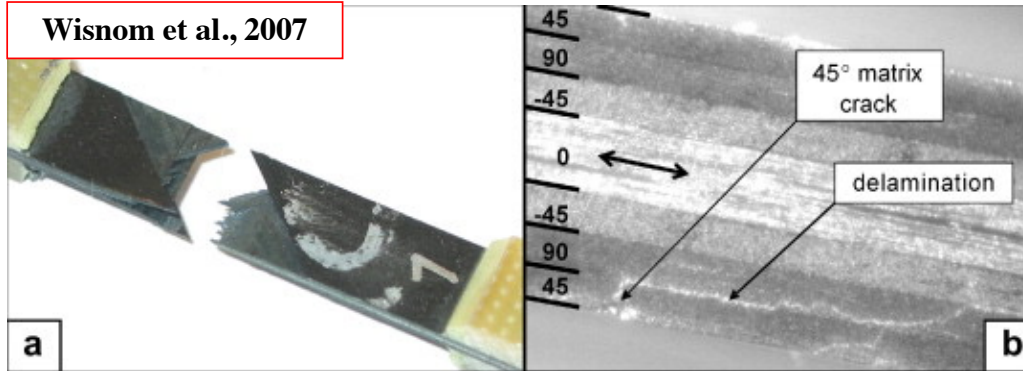


buckling
crushing

🕒 Understanding the fracture behavior requires a multiscale approach for characterization and modeling to account for the interaction between length scales.



Intralaminar failure (fiber failure, matrix cracking, interface decohesion)



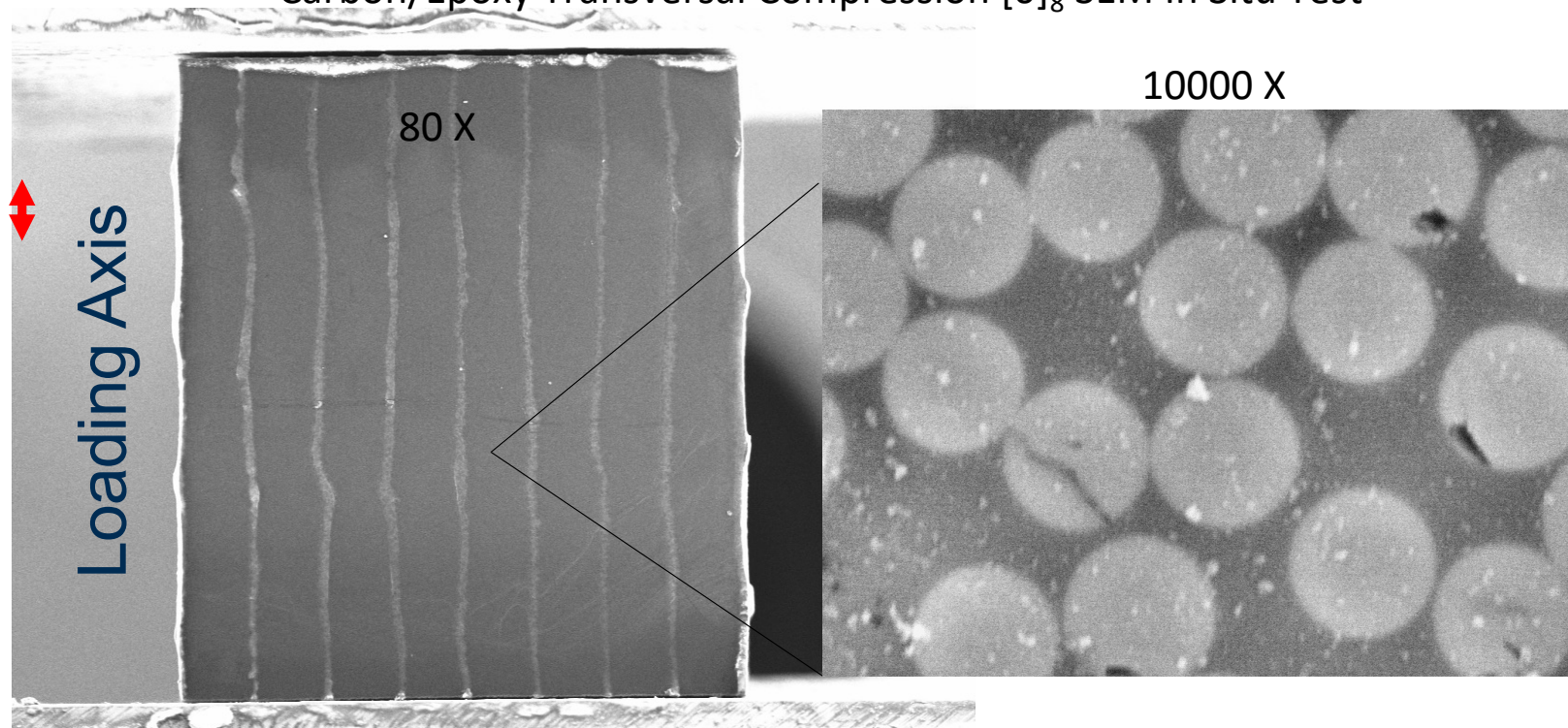
Non linearity (geometry and material)

Interlaminar failure (delamination)



Damage triggered at the microscale

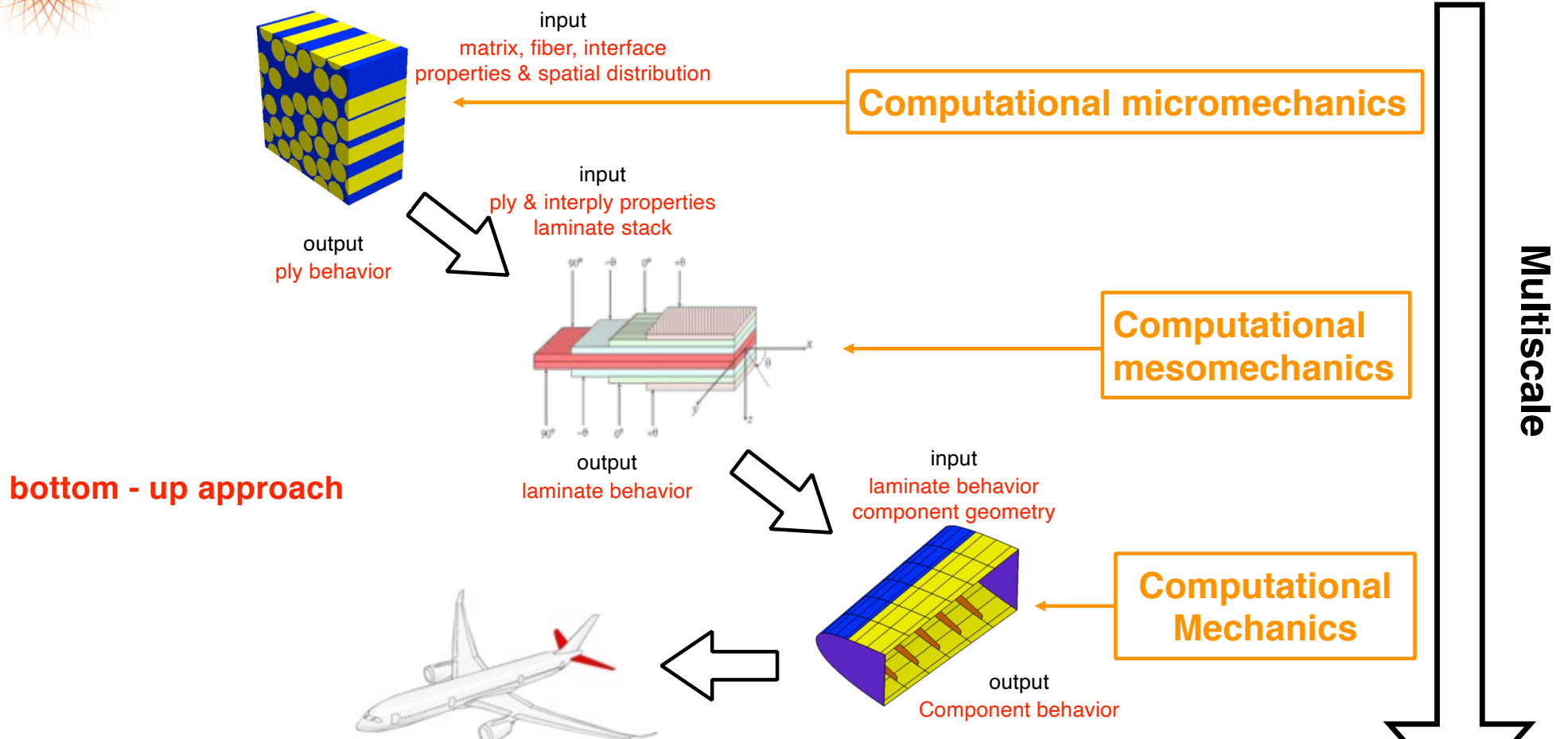
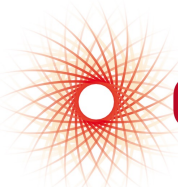
Carbon/Epoxy Transversal Compression $[0]_8$ SEM In Situ Test



Plastic shear banding and failure

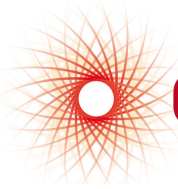
Fiber debonding + matrix cracking



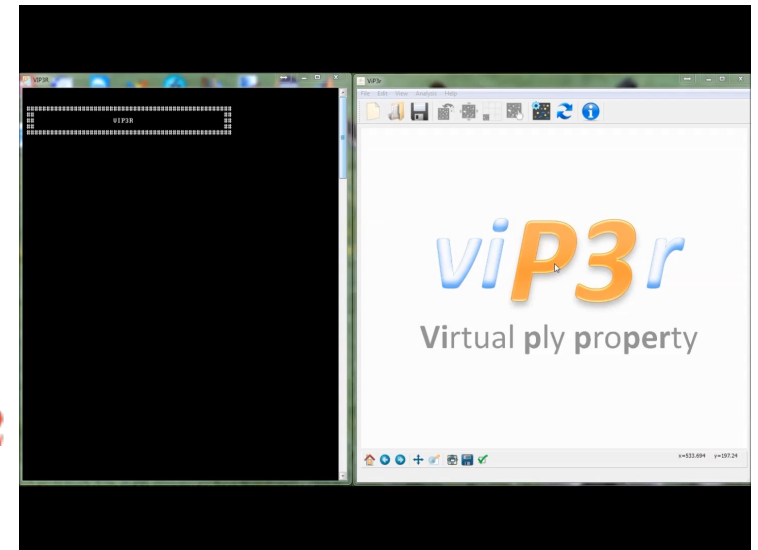
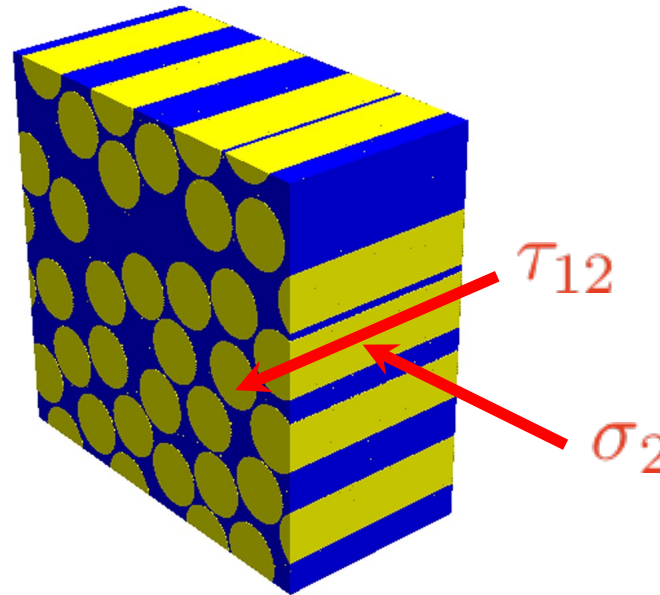
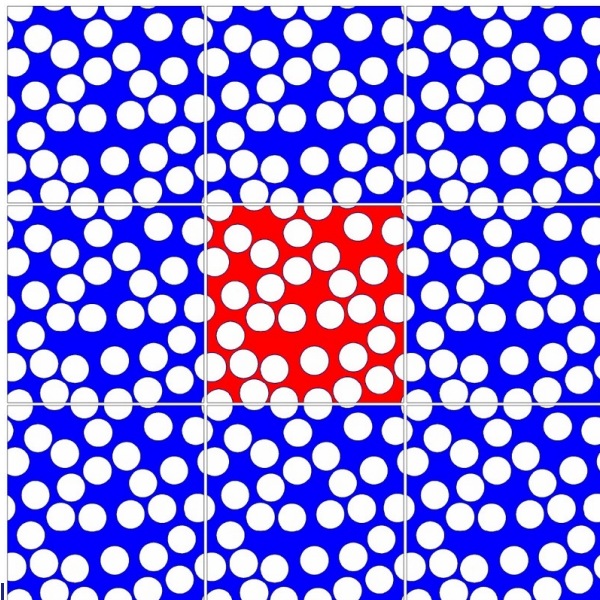


This project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No 101007022.

Llorca, J, Gonzalez, C, Molina-Aldareguía, J M, Segurado, J, Seltzer, R, Sket, F, Rodriguez, M, Sadaba, S, Munoz, R, Canal, L P, Multiscale modeling of composite materials: a roadmap towards virtual testing, *Advanced materials*, 23, 5130-47, 2011



- Fiber centers are generated randomly and sequentially to reach target volume fraction in a square RVE. A periodic microstructure is formed by an indefinite translation of the RVE.
- A 3D parallelepiped containing a random and homogeneous dispersion of fibers is obtained by extruding the 2D RVE along the fiber axis.
- Periodic boundary conditions are imposed between opposite faces of the RVE (jig-saw puzzle)



👤 Fibers behave as elastic solids: transverse isotropic (carbon fibers) and isotropic (glass fibers)

FIBER MODEL

👤 Interface decohesion is accounted for using a cohesive crack model

INTERFACE MODEL

- Onset of damage $\max\left\{\frac{\langle t_n \rangle}{N}, \frac{t_s}{S}\right\} = 1$

- Traction-separation law

$$t_n = (1 - d)K\delta_n \quad \text{if } \delta_n > 0$$

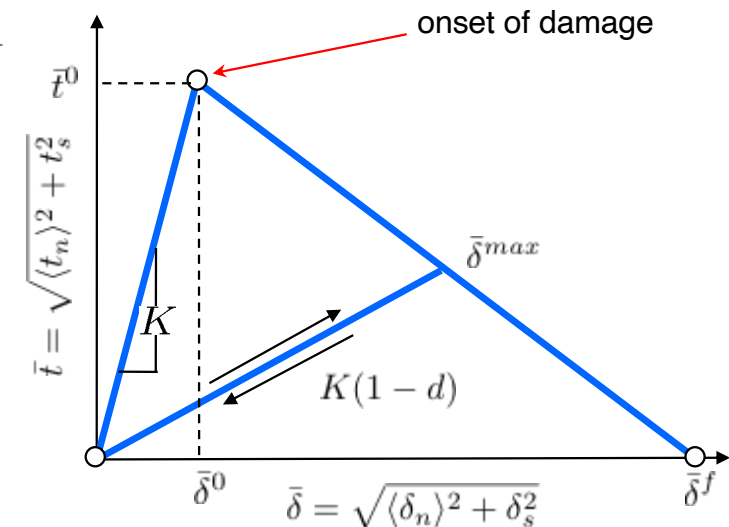
$$t_n = K\delta_n \quad \text{if } \delta_n \leq 0$$

$$t_s = (1 - d)K\delta_s$$

- Evolution of the damage parameter

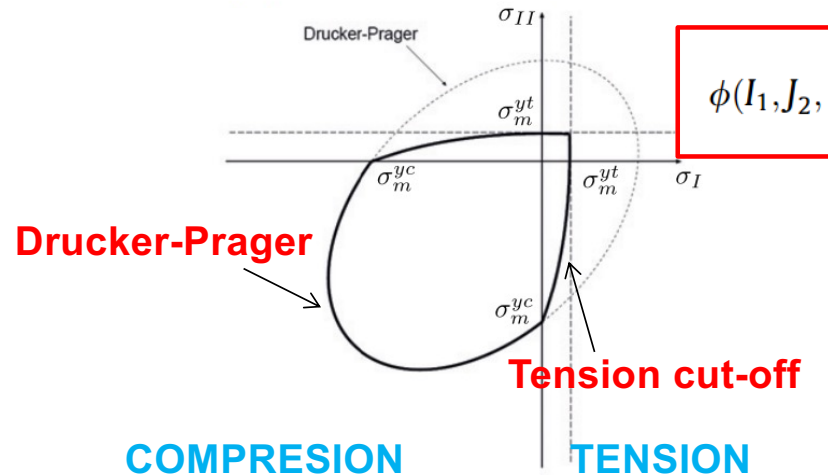
$$d = \frac{\bar{\delta}^f (\bar{\delta}^{max} - \bar{\delta}^0)}{\bar{\delta}^{max} (\bar{\delta}^f - \bar{\delta}^0)}$$

- Interface fracture energy $G_F = \frac{1}{2} \bar{t}^0 \bar{\delta}^f$

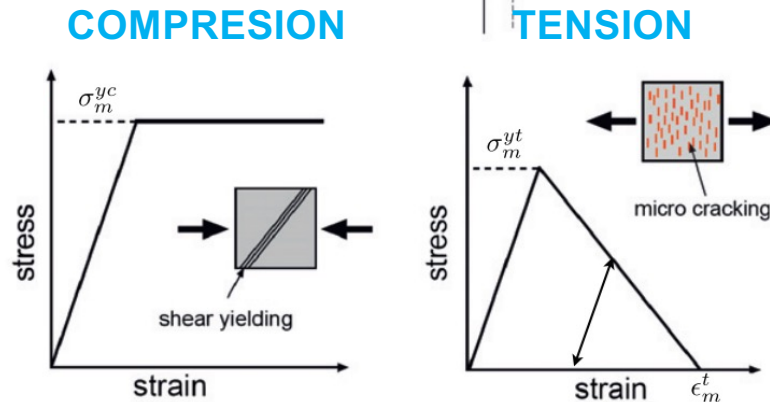


The epoxy matrix followed a modification of Drucker-Prager plasticity (Lubliner) to account for brittle fracture in tension and shear yielding

MATRIX MODEL

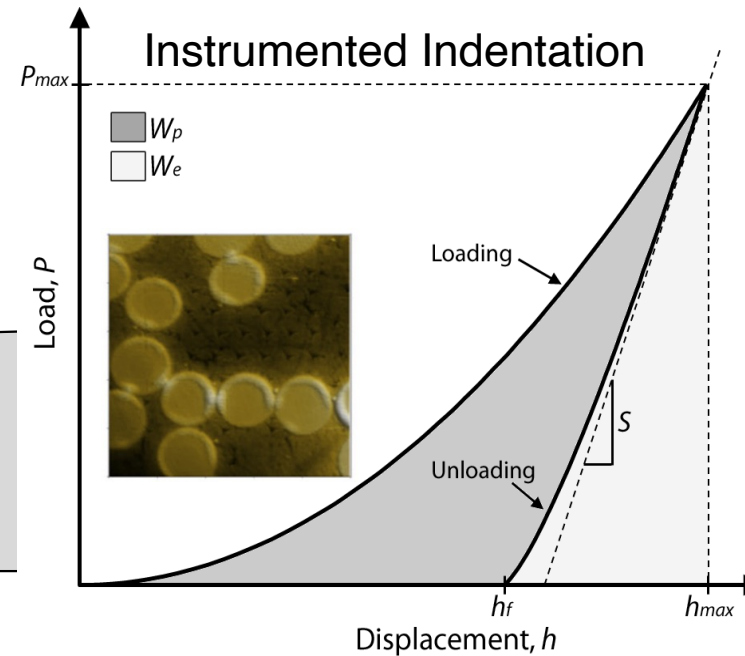
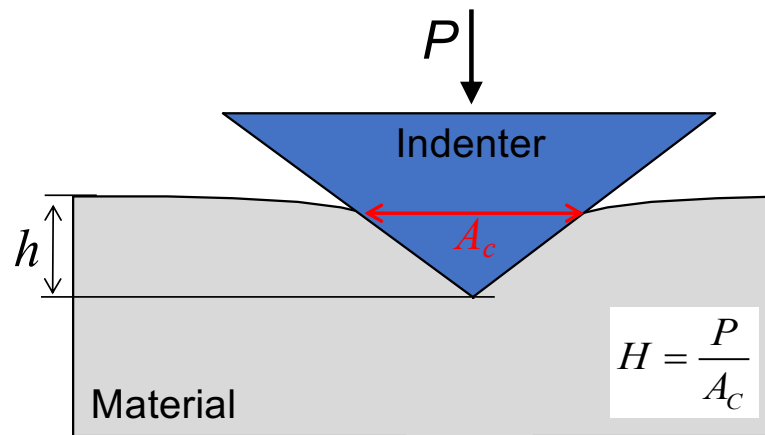


$$\phi(I_1, J_2, \sigma_I, \beta, \alpha) = \frac{1}{1 - \alpha} \left(\sqrt{\frac{3J_2}{2}} + \frac{I_1}{3}\alpha + \beta \langle \sigma_I \rangle \right) - \sigma_m^{yc} = 0$$

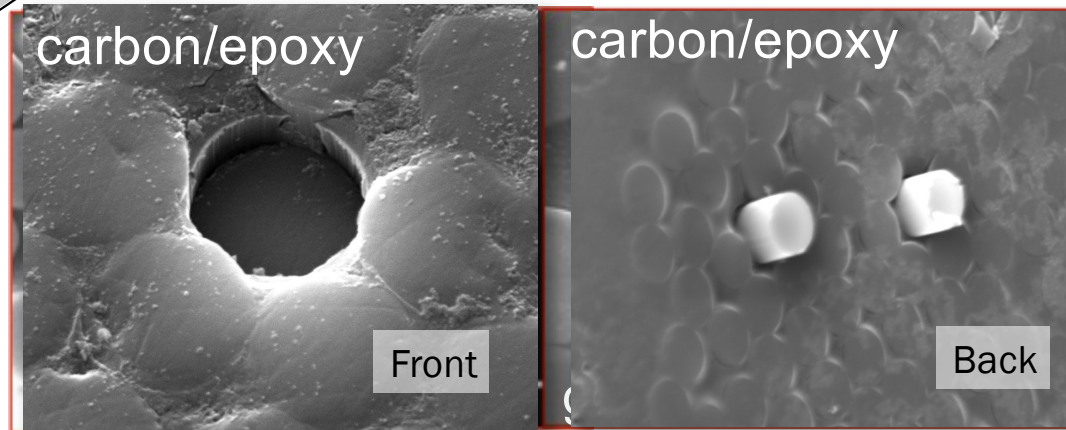
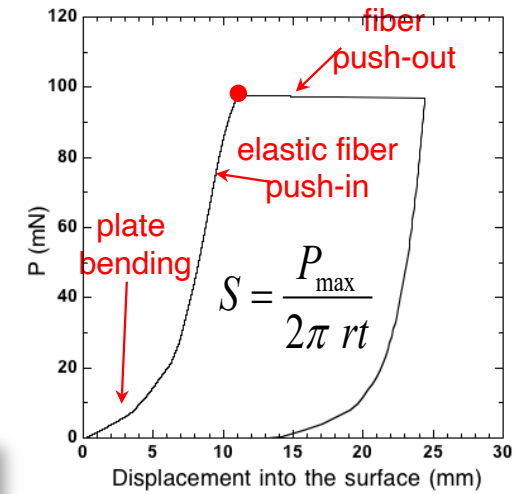
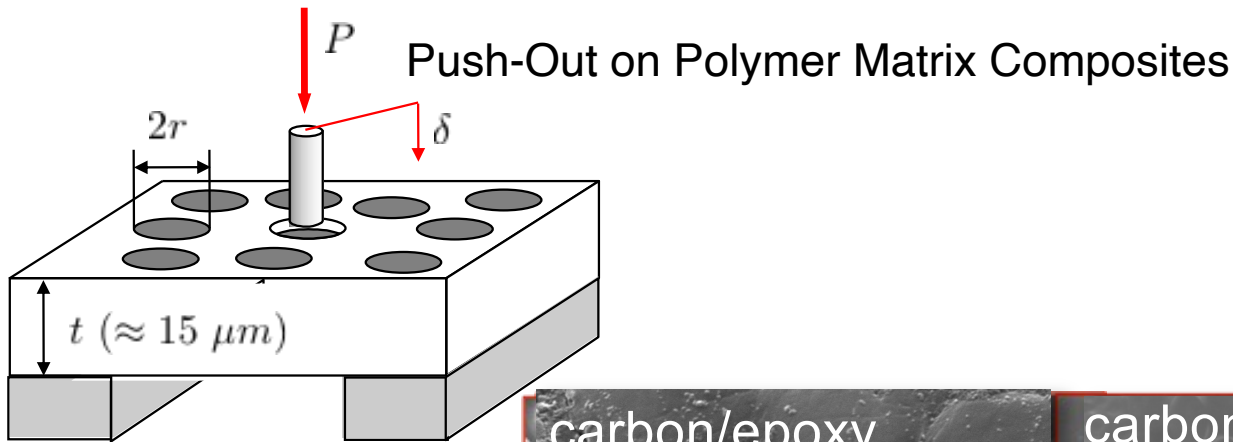


- α • Pressure sensitivity
 - σ_m^{yc} • Compression yield stress
 - σ_m^{yt} • Tension cut-off
 - G_m^f • Fracture energy for softening
- $$\beta = \frac{\sigma_m^{yc}}{\sigma_m^{yt}} (1 - \alpha) - (1 + \alpha)$$

MATRIX PROPERTIES

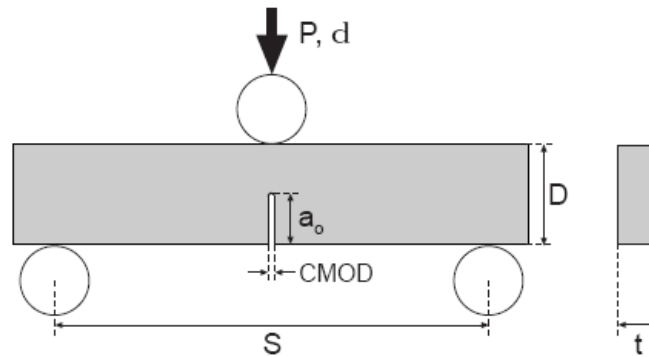


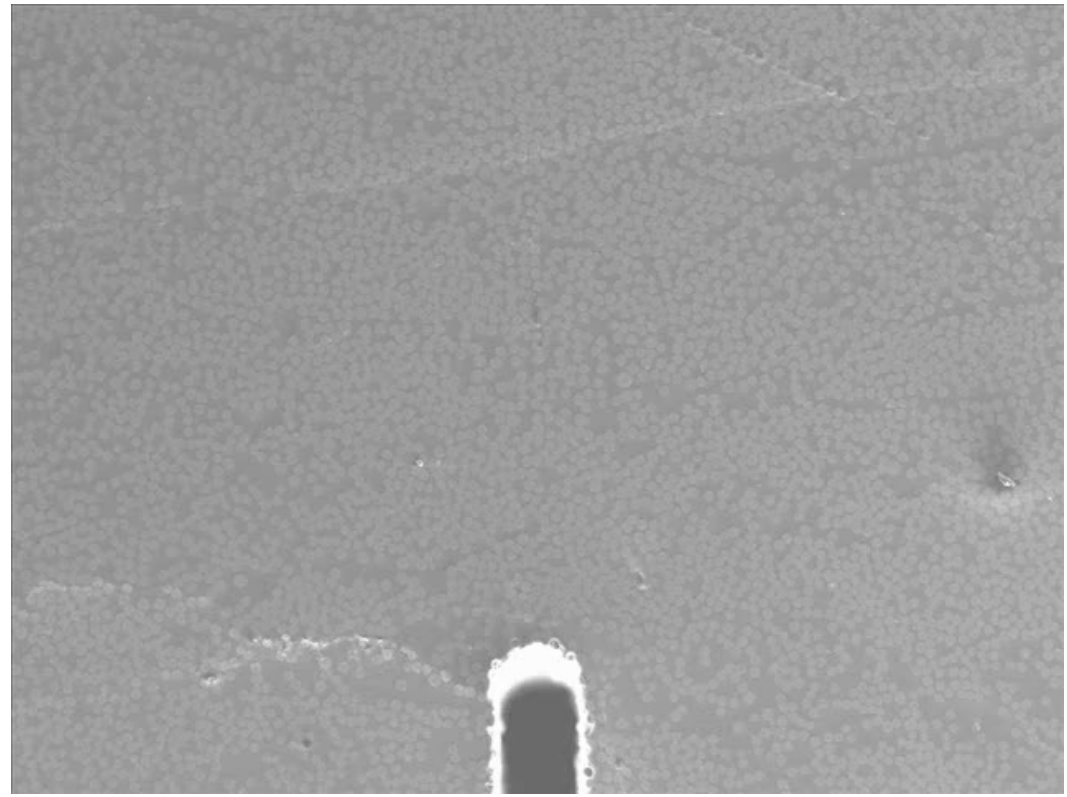
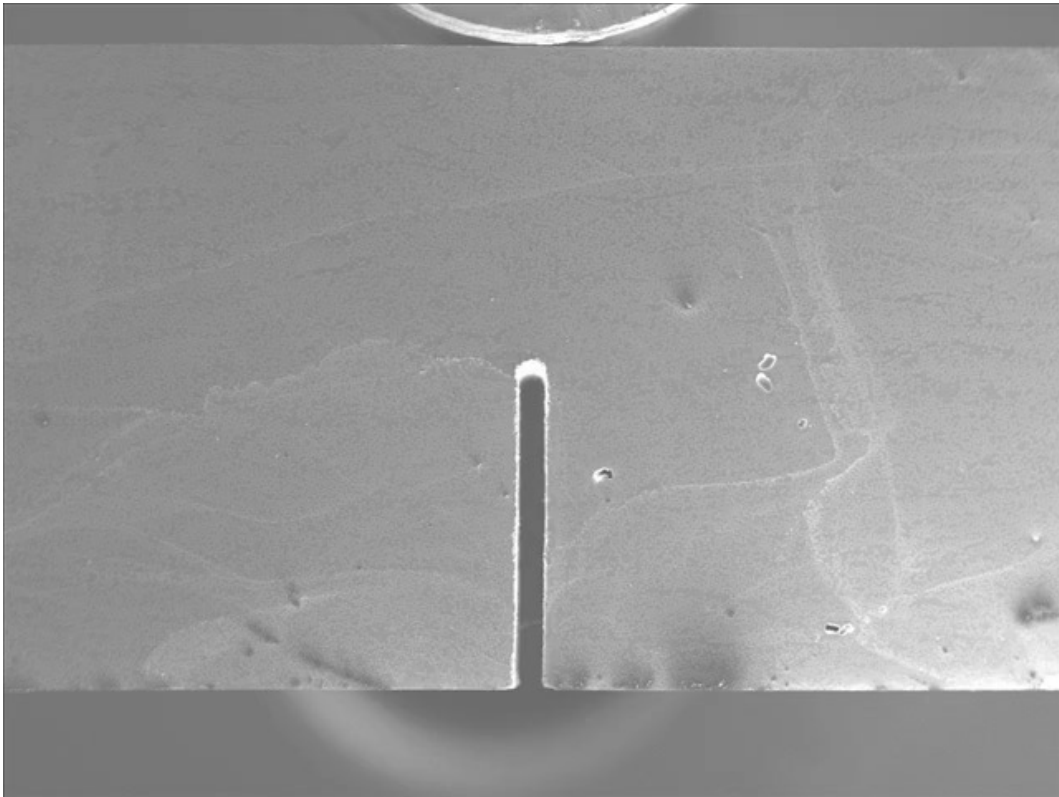
- 🕒 Load P vs. depth h of the indenter is monitored
- 🕒 Hardness H is dependent on the yield strength σ_m^{yc}
- 🕒 Real and apparent contact areas differences (pile-up and sink-in)
- 🕒 FEM detailed model of indentation in pressure dependent materials
- 🕒 Drucker Prager solids σ_m^{yc} and α



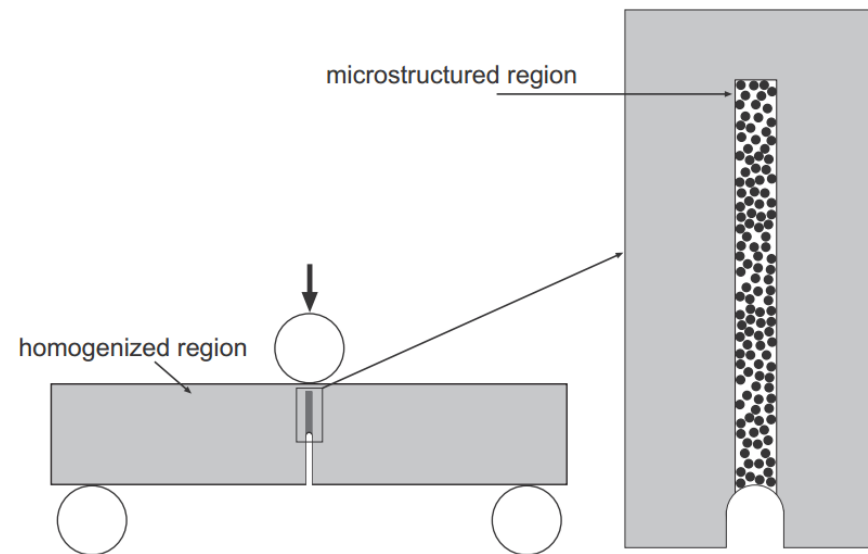
THREE POINT BENDING

- TPB notched specimens
- $a_0/D \approx 0.5$ for stable crack growth
- On-line detailed inspection of failure micromechanisms

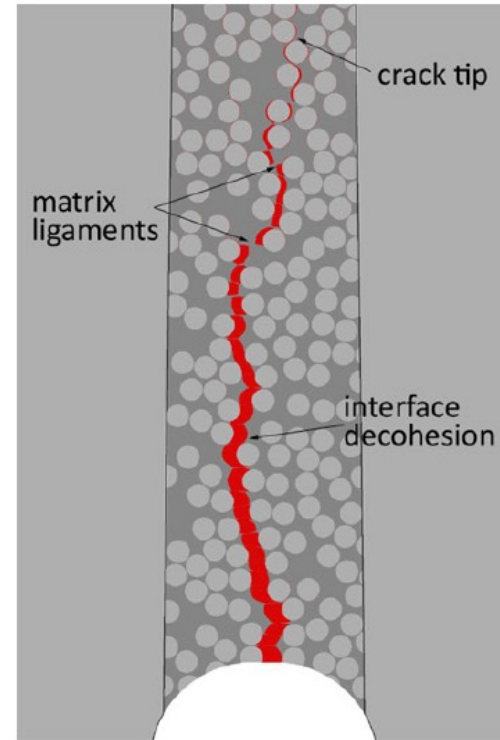
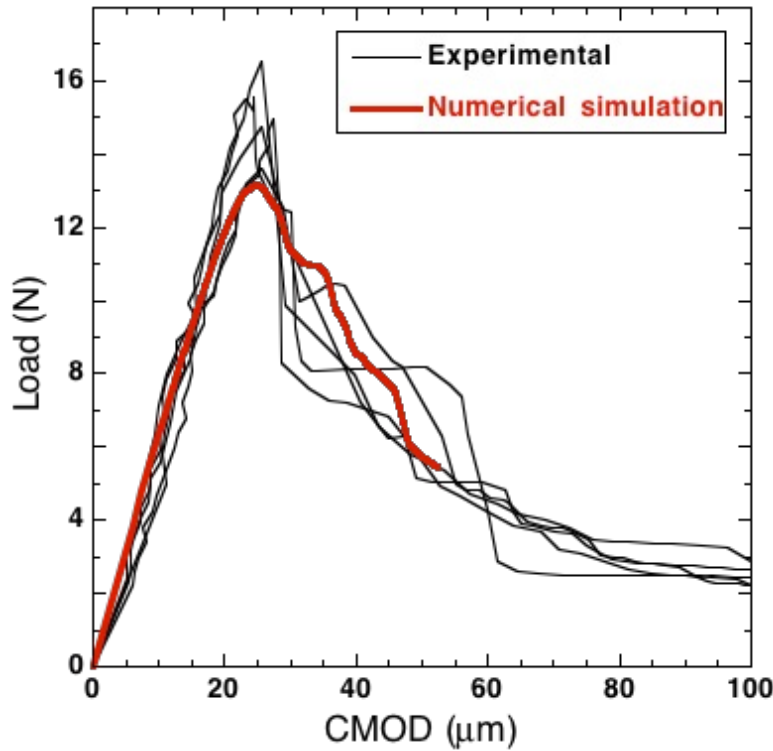




Simulations were carried out with Abaqus/standard under plane strain conditions within the framework of an embedded cell model.

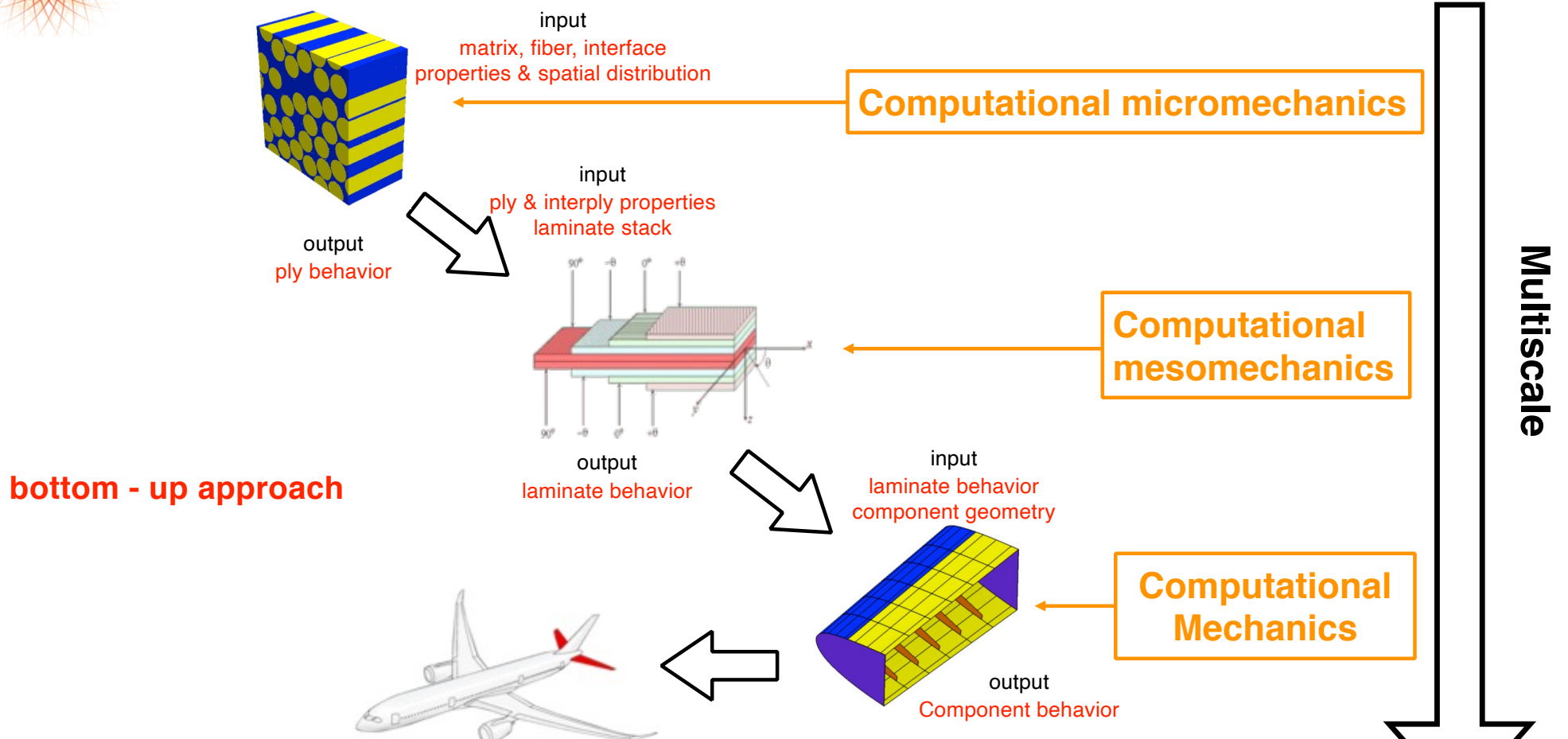
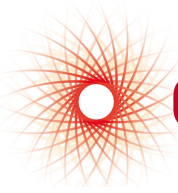


- The elastic constants of the transversely isotropic homogenized composite were obtained from those of the matrix and fibers using Mori-Tanaka
- Discretization was carried out with quadratic triangles (CPE6M)
- Interfaces were discretized with linear triangles (COH2D)



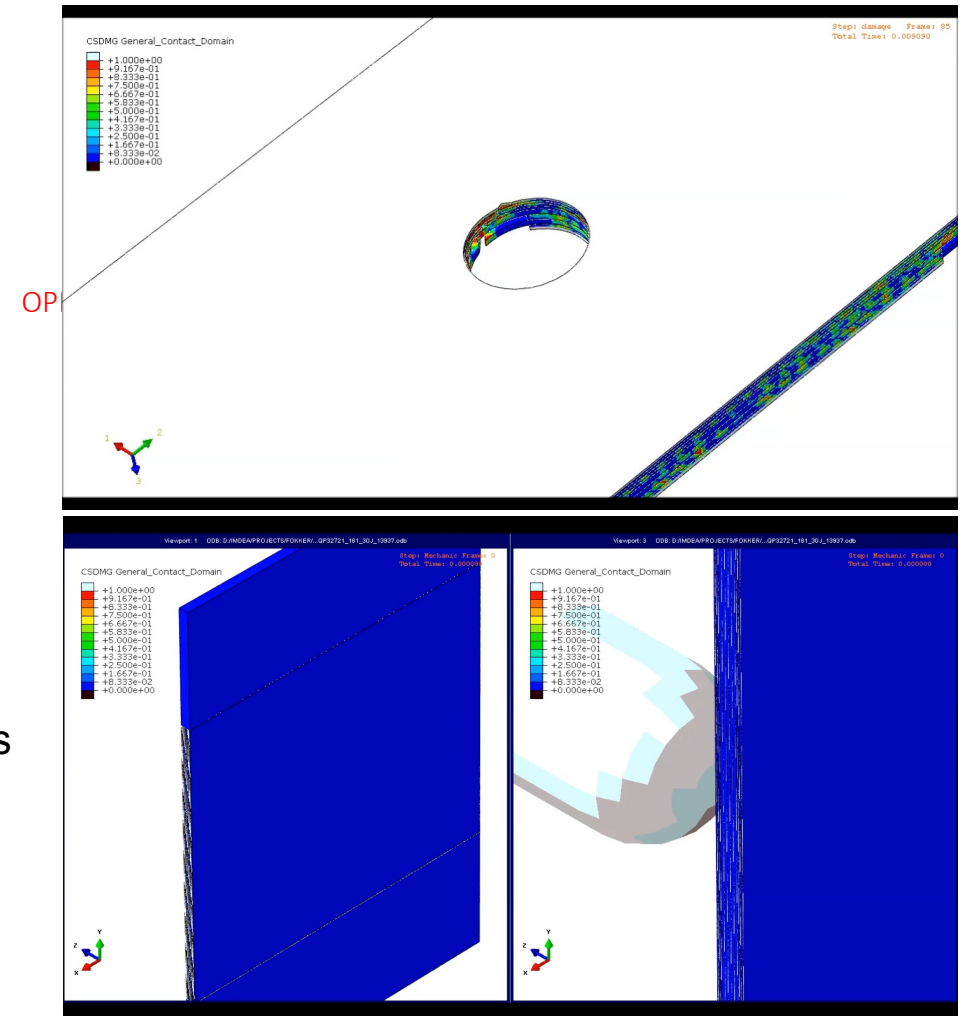
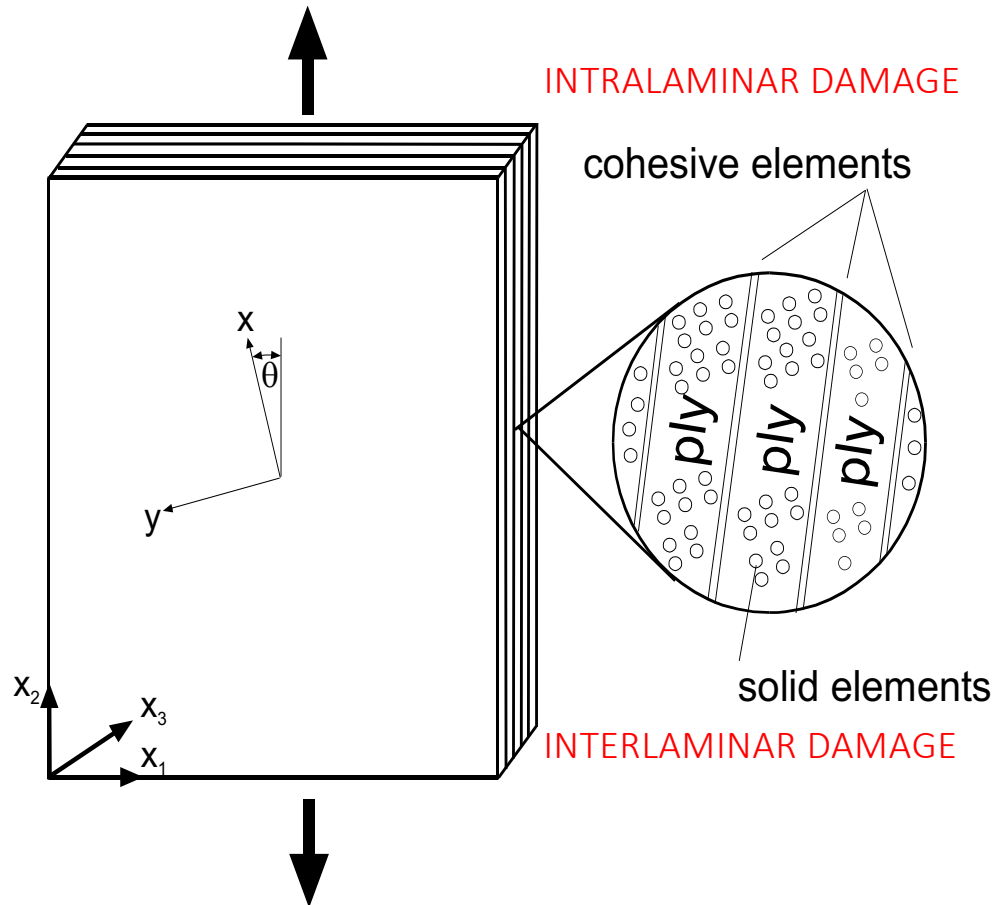
👤 The numerical model is able to reproduce accurately the macroscopic P- CMOD curves as well as the microscopic failure mechanisms.

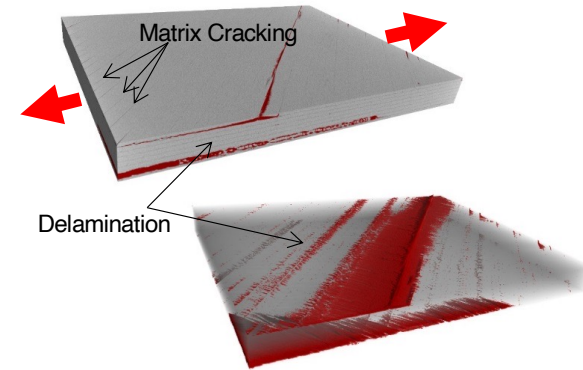
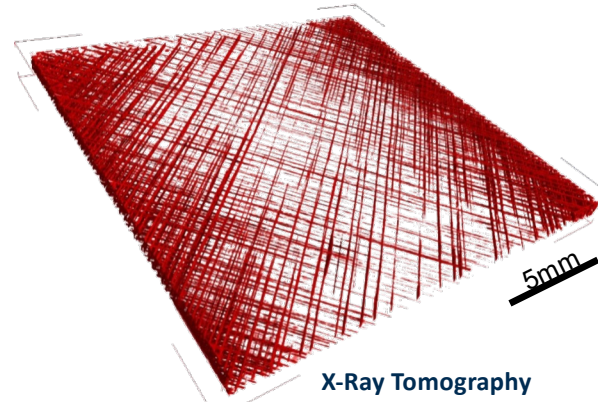
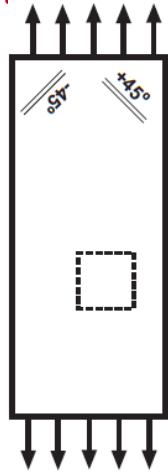




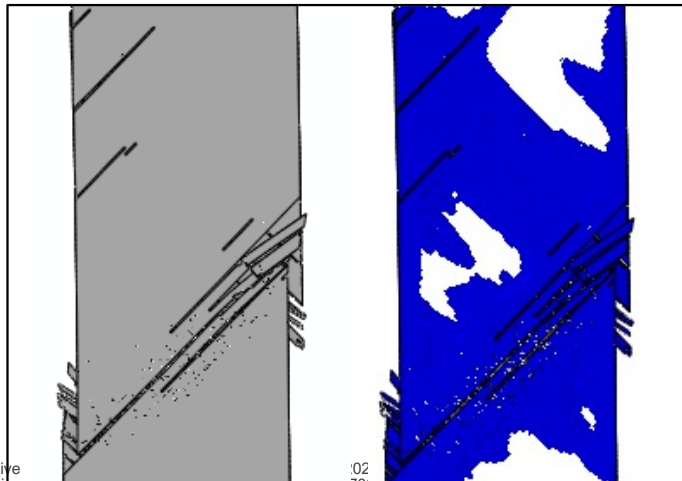
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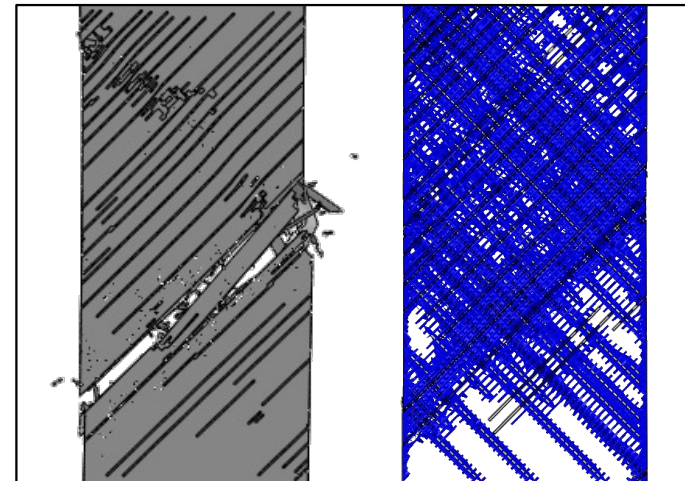




IN-PLANE SHEAR



NO residual thermal stresses



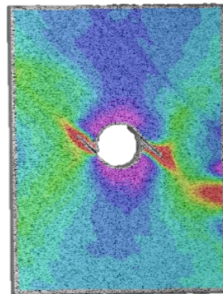
WITH residual thermal stresses



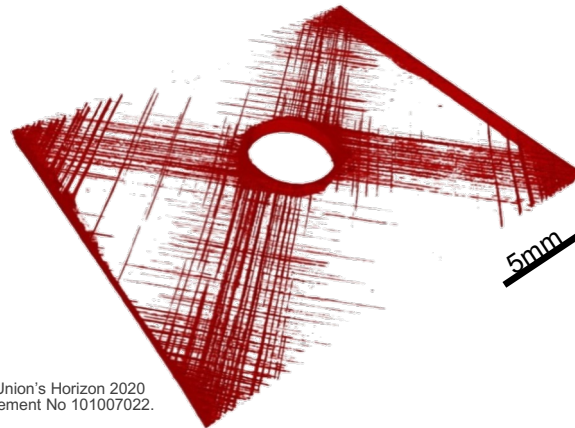
OPEN-HOLE TENSION



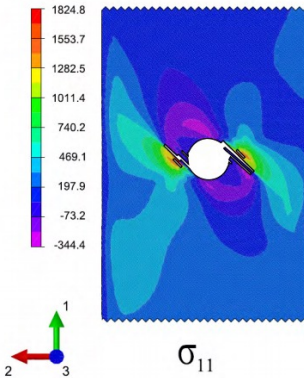
Digital Image Correlation



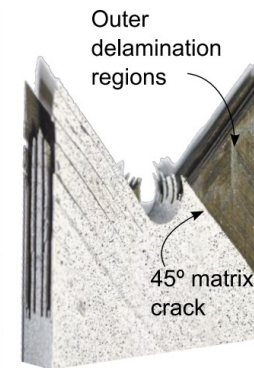
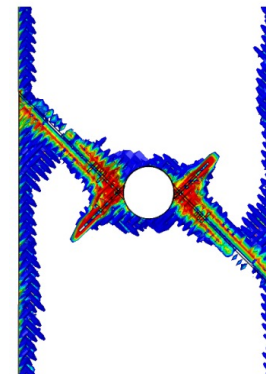
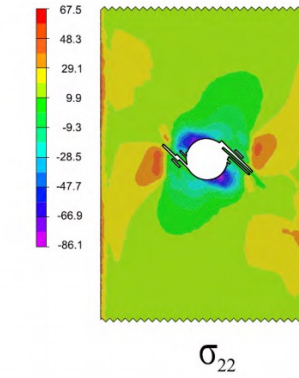
X-Ray Tomography



S, S11
(Avg: 75%)

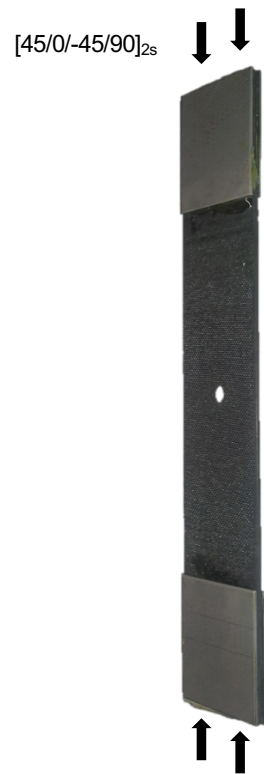


S, S22
(Avg: 75%)

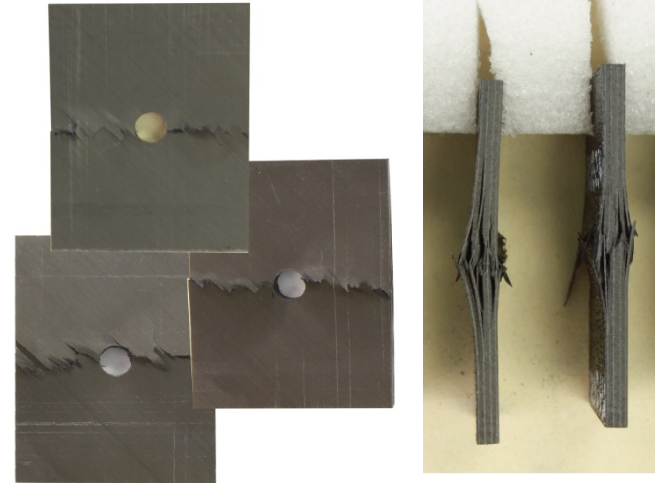




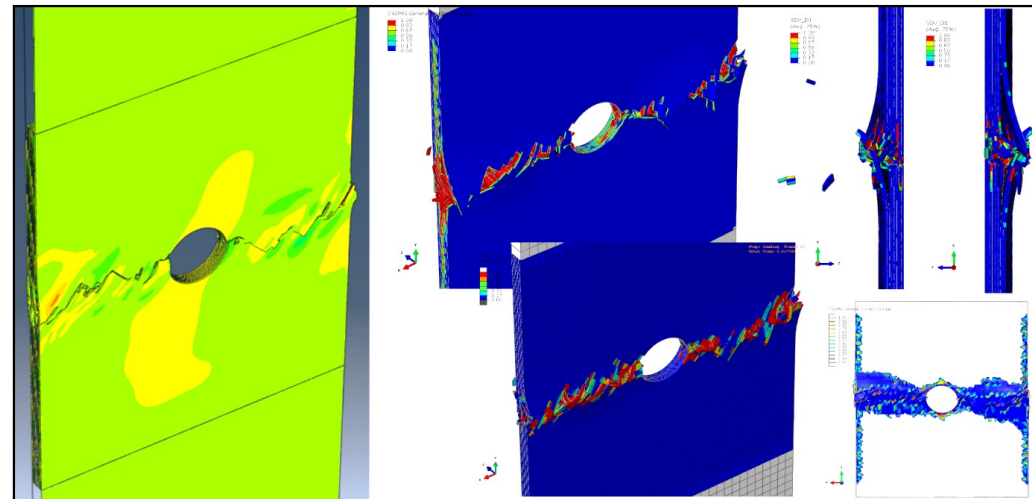
OPEN-HOLE COMPRESSION



Experiments



FEM



AS4/8552

Stiff
Medium - shear dominated

Soft

Soft

Laminate	Test strength (MPa)	Predicted strength (MPa)	Error (%)
In-Plane Shear	97.5	94	-3.6
Plain Tension - QI	663	673	1.5
Plain Compression - QI	540	570	5.6
Plain Tension - [50/40/10]	1076	1040	-3.3
Plain Compression - [50/40/10]	831	757	-8.9
Plain Tension - [10/80/10]	436	458	4.9
Plain Compression - [10/80/10]	391	404	3.4
Plain Tension - [30/40/30]	761	770	1.1
Plain Compression - [30/40/30]	540	603	11.5
Open-Hole Tension - QI	371	401	7.9
Open-Hole Compression - QI	304	306	0.6
Open-Hole Tension - [30/40/30]	446	402	-9.8
Open-Hole Compression - [30/40/30]	311	299	-3.9



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low

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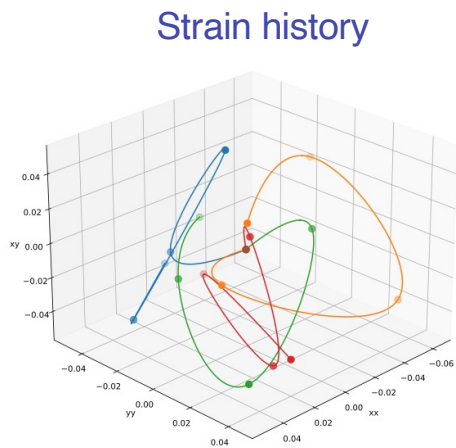


advantages

- 🕒 Close to the actual failure mechanisms
- 🕒 Influence of microstructure and constituents properties
- 🕒 Influence of the lay-up, stacking sequence, etc.

disadvantages

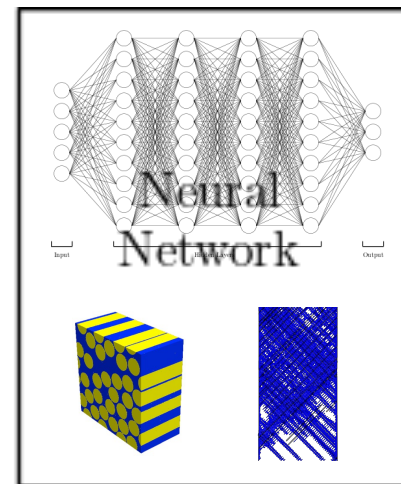
- 🕒 Time consuming
- 🕒 Requires expertise on modelling



ϵ_{11}
 ϵ_{22}
 ϵ_{33}
 ϵ_{23}
 ϵ_{13}
 ϵ_{12}



Digital twin

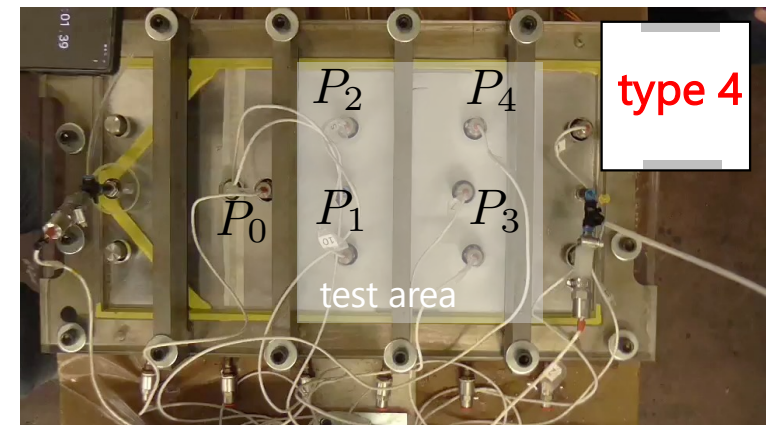
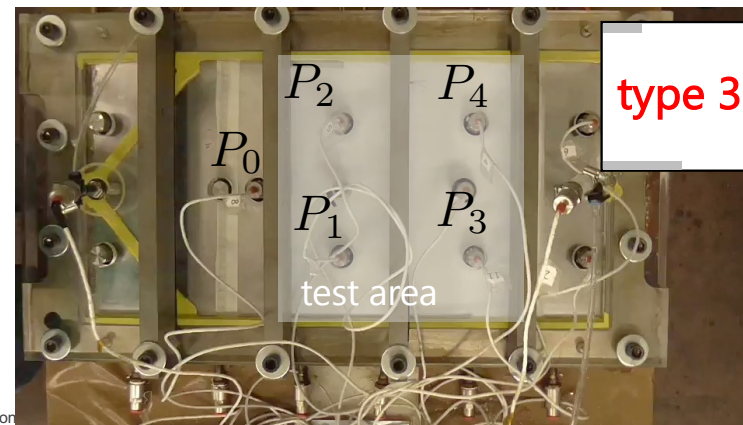
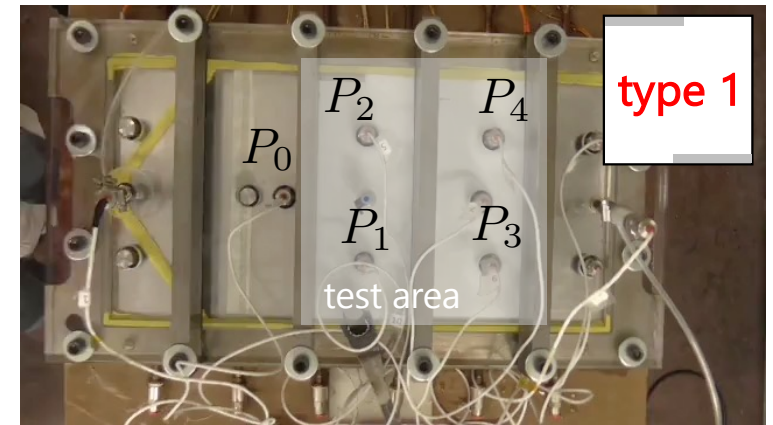
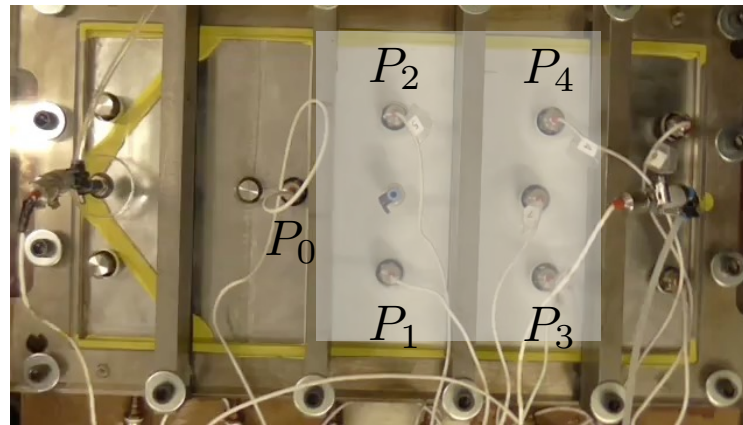


Stress results

σ_{11}
 σ_{22}
 σ_{33}
 σ_{23}
 σ_{13}
 σ_{12}

Forward Mechanical Problem Micro & Mesomechanics

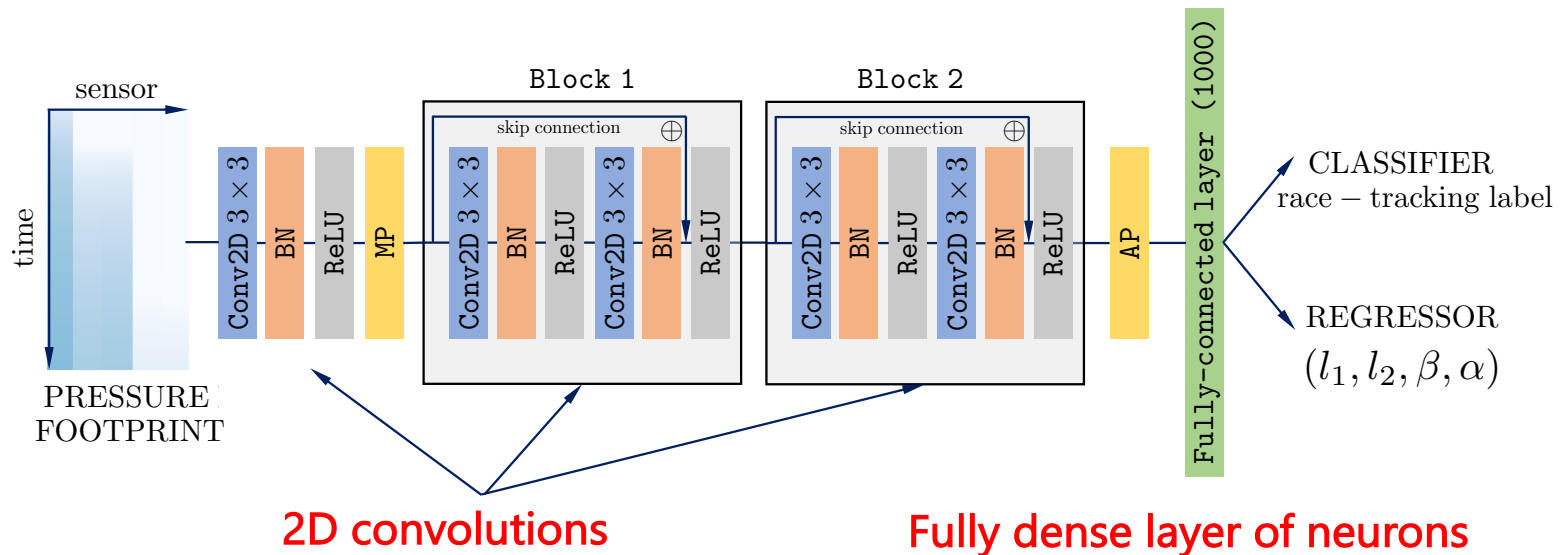
no race-tracking



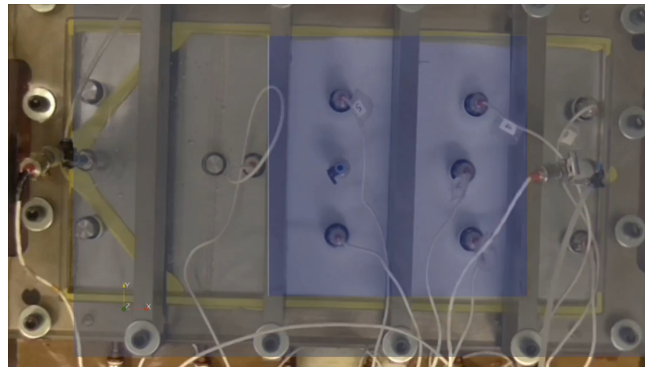
TRAINING THE NETWORK



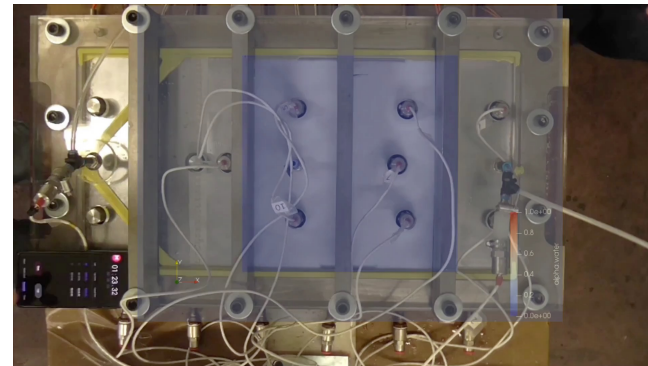
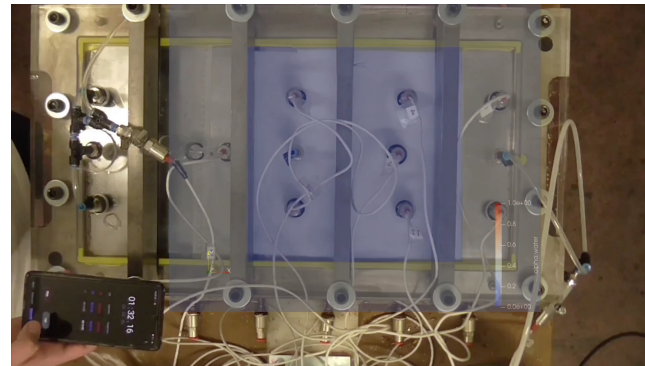
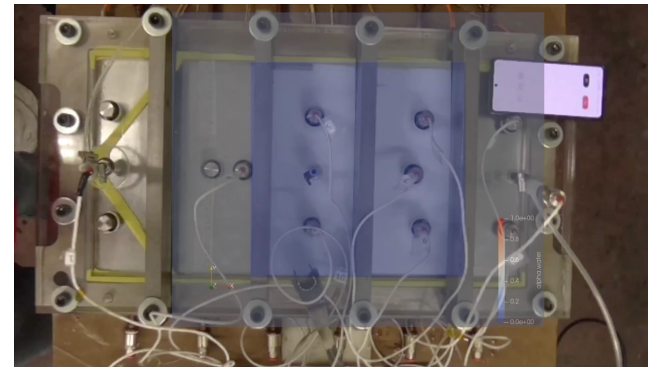
- 2000 OpenFoam simulations as data-sets
- Deep Residual Learning for Image Recognition ResNet*
- Classification optimization based on CE loss (cross-entropy)
- Regression optimization based on MSE loss (mean square error)



no RT



type 1



type 3

type 4

SIM-EXP OVERLAID VIDEOS



Thank you for your attention

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This project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No 101007022.