

**19<sup>th</sup> X-ray & CT Forum**

**September 10<sup>th</sup>-12<sup>th</sup> 2024  
Hamburg (Germany)**

## **ADVANCES IN XCT: from the auxiliary role of AI to the study of in situ processes**

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Baker Hughes Confidential



December 2019 – December 2023



# Where are we?



Ubicación de Getafe en España.



Ubicación de Getafe en Comunidad de Madrid.



espormadrid.es



- **Public research centre since 2007.**
- **16 research groups.**
- **120 JCR published papers per year.**
- **> 150 people.**
- **> 70 R&D private contracts.**

# In-situ processing and mechanical characterization of materials group

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# The X-Ray Laboratory



- **Coordinated by the “In-situ processing and mechanical characterization of materials” group.**
  - **Advanced characterization of materials, including microstructural, chemical, and crystallographic information on various scales of magnitude using different techniques.**
- **Key laboratory in a research line focused on multi-scale characterization of materials and processes.**
- **We are going several times per year to different European synchrotrons (ESRF, SLS, DESY, BESSY, ALBA).**

# The X-Ray Laboratory



Phoenix (GE) Nanotom  
160NF – Lab tomograph  
(absorption)

phoenix



Zeiss Xradia 620 Versa – Lab  
tomograph (absorption,  
phase & diffraction)

zeiss.com



PANalytical Empyrean – Lab  
diffractometer (diffraction & reflection)

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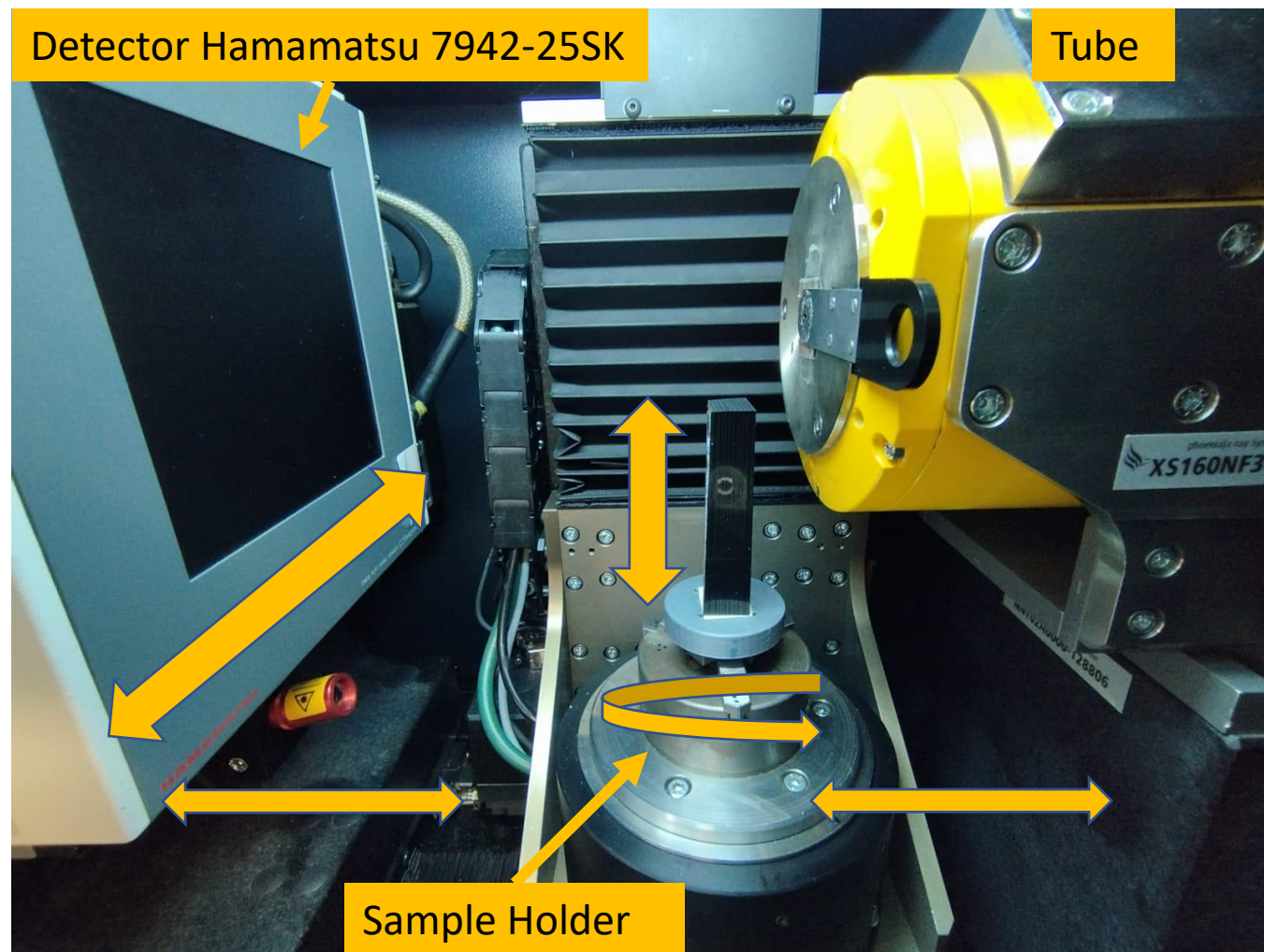
malvernpanalytical.com

# The X-Ray Laboratory



suplitec-ndt.com

- Max voltage: 160 kV.
- Targets: Mo and W.
- Voxel size: from 30 to 1  $\mu\text{m}/\text{px}$ .
- Detector area: 2300 $\times$ 2300 (pixel side 50  $\mu\text{m}$ ).
- Three virtual detectors.
- Up to 9 radiographs per second.

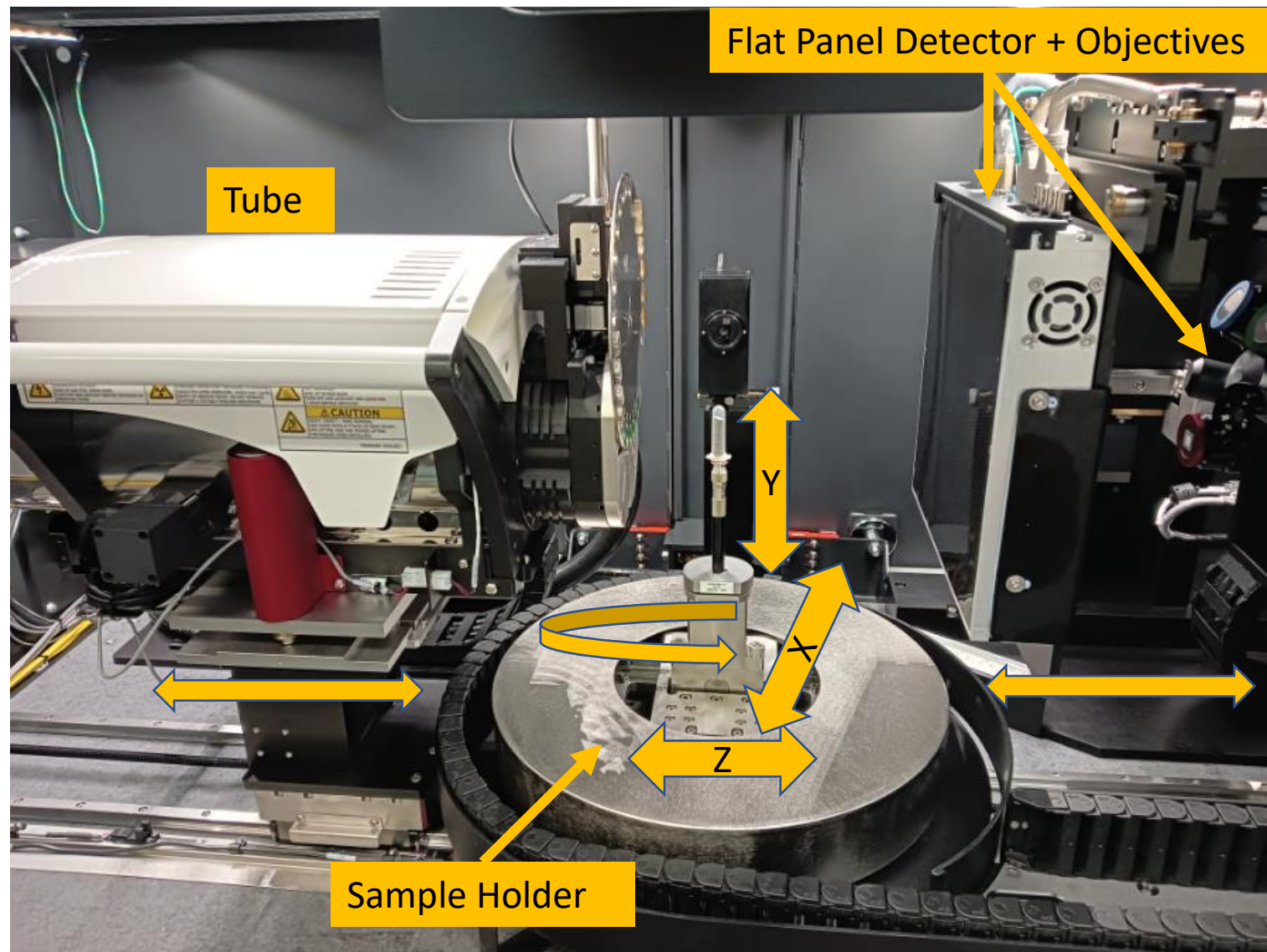


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# The X-Ray Laboratory



- Max voltage: 160 kV.
- Target: W.
- Voxel size: from 57 to 0.3  $\mu\text{m}/\text{px}$ .
- Detector area: 3072 $\times$ 1944 (FPX, 75  $\mu\text{m}$ ) and 2048 $\times$ 2048 (CCD, 14  $\mu\text{m}$ ).
- Objectives: 0.4X, 4X, 20X, 40X.
- DCT option is available.

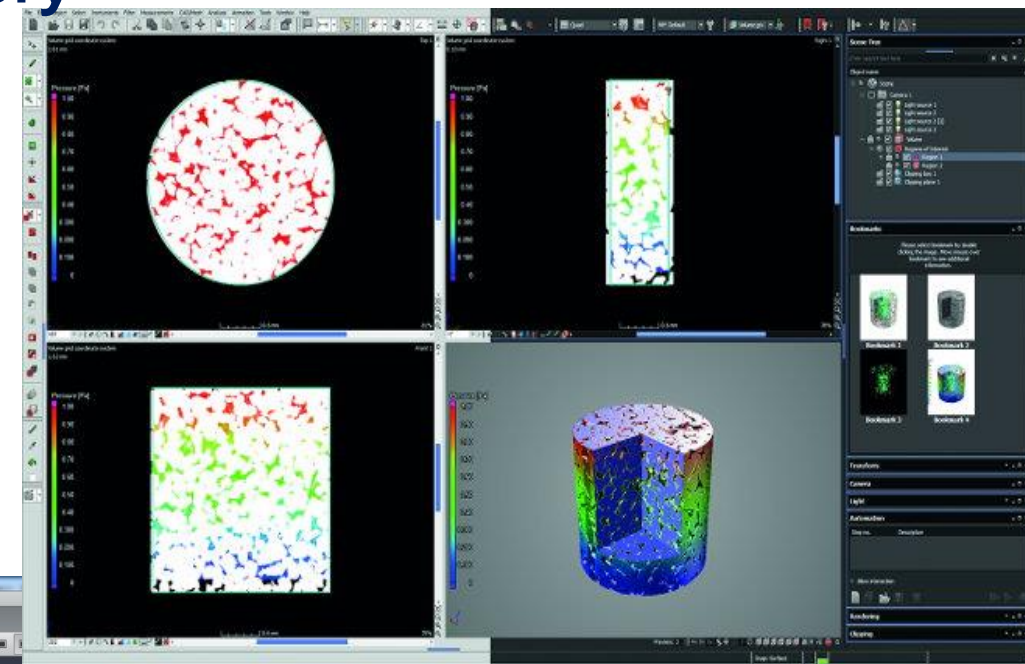




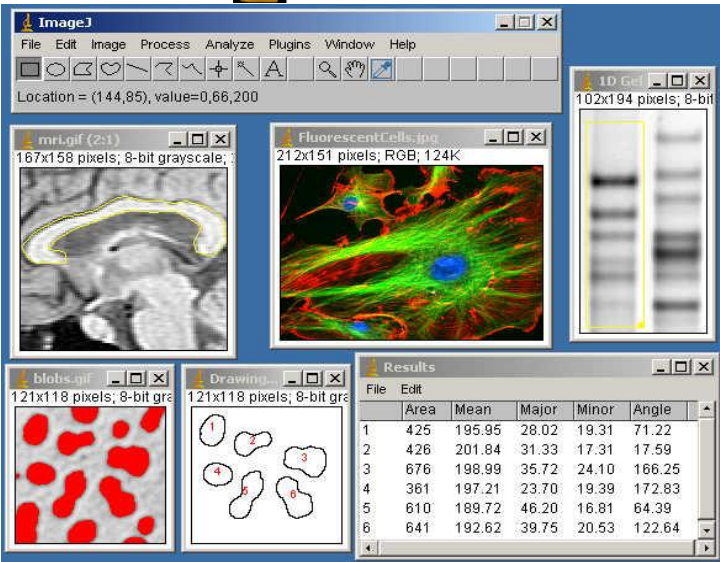


**ImageJ**  
Image Processing & Analysis in Java

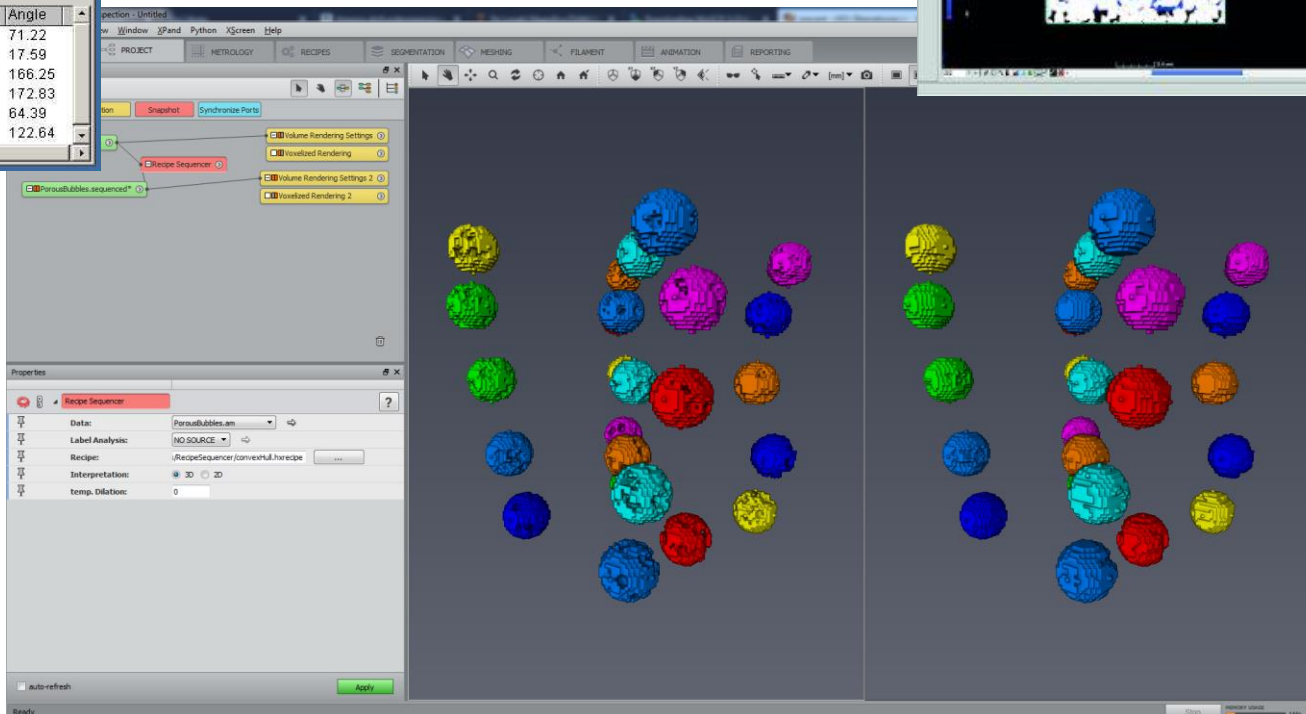
# The X-Ray Laboratory



Avizo®



University of Arizona

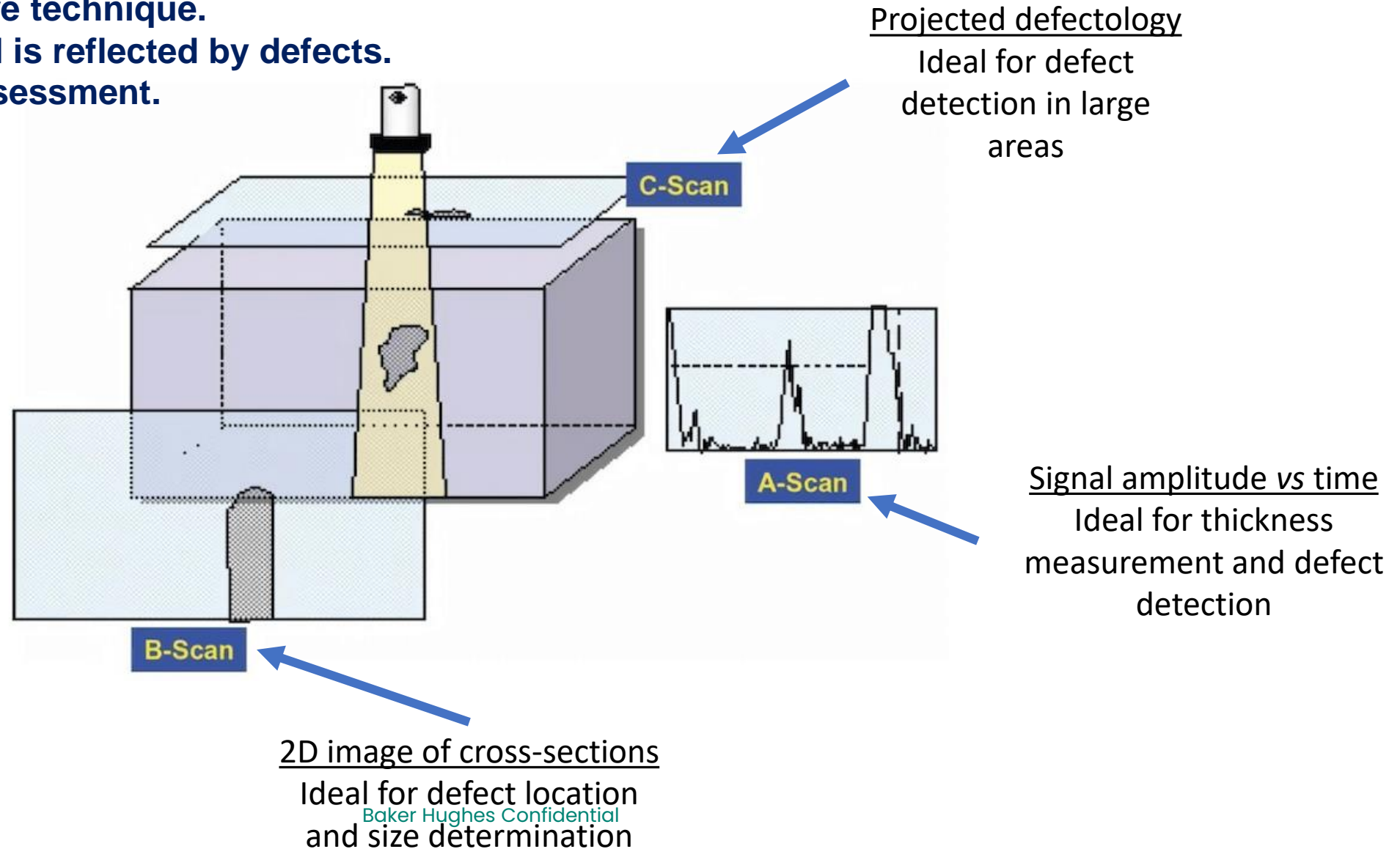


ndt.net

# Combining XCT with ultrasonic tests (UT) for CFRP composites

# XCT vs UT

- **UT is a non-destructive technique.**
- **High frequency sound is reflected by defects.**
- **Applied for quality assessment.**

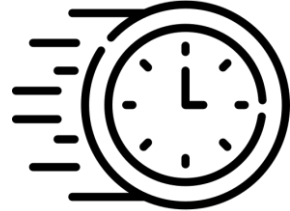


# XCT vs UT

- Ultrasound Testing

- X-Ray Computed Tomography

Fast



Low detail



Slow



Detailed



# XCT vs UT

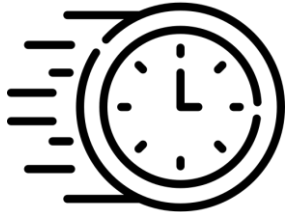
- Our goal is to combine the scan speed of UT with the quality of XCT.
- Deep learning approach is the most convenient.

• Ultrasound Testing

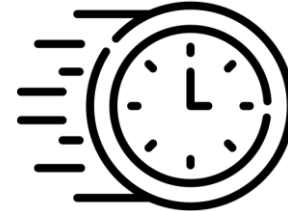


• X-Ray Computed Tomography

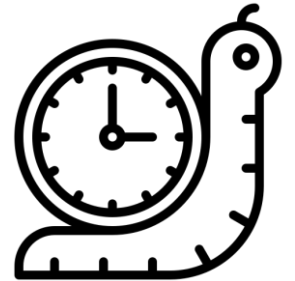
Fast



Fast



Slow



Low detail



Detailed



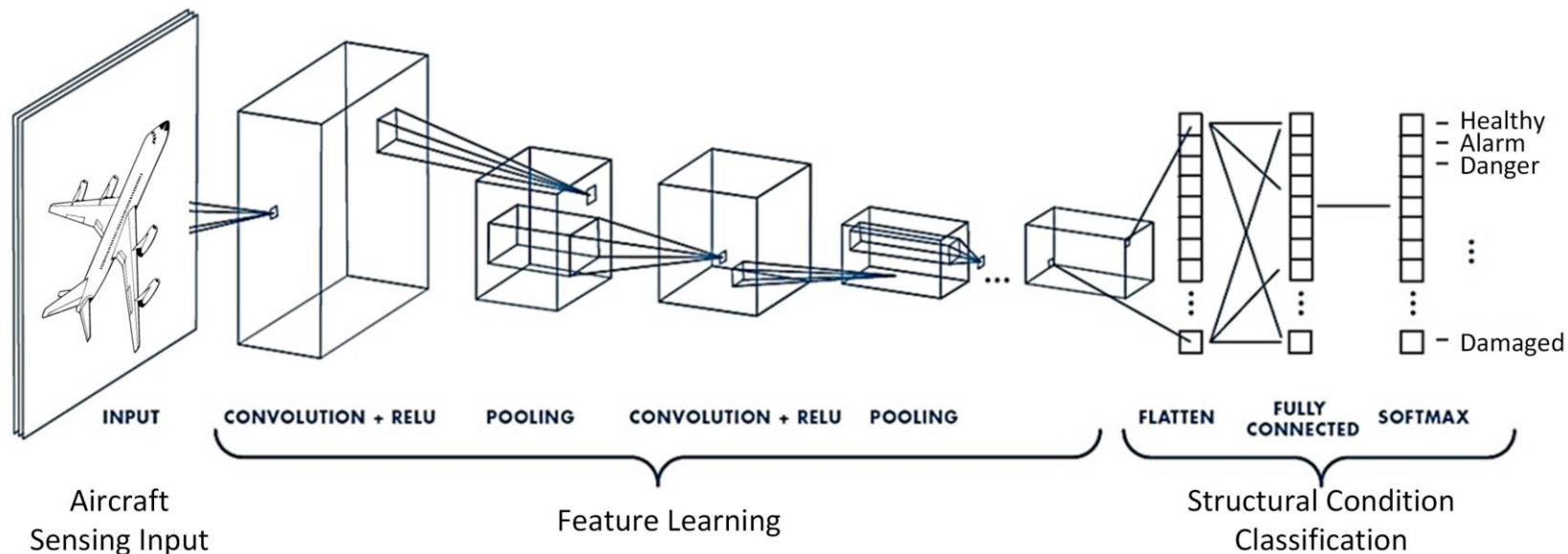
Detailed



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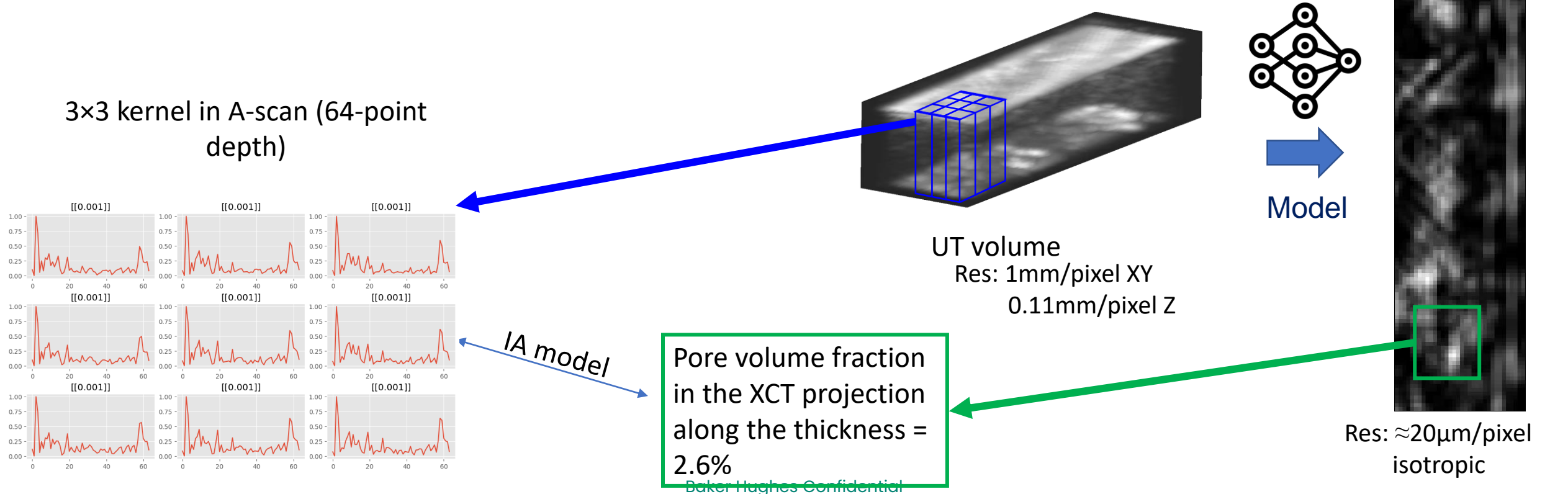
# XCT vs UT

- **UT output is a 3D volume of information (X, Y, time window).**
- **Normally, ultrasound volume is converted into C-scans.**
- **We can think about that an ultrasound volume is a stack of A-scans.**
- **This information can be treated using a Convolutional Neural Network.**



# XCT vs UT

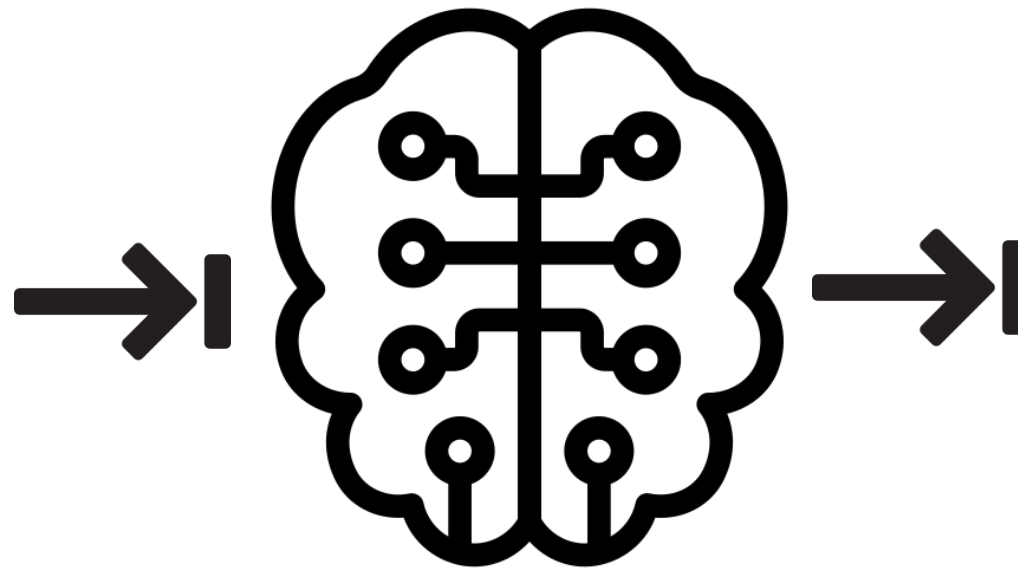
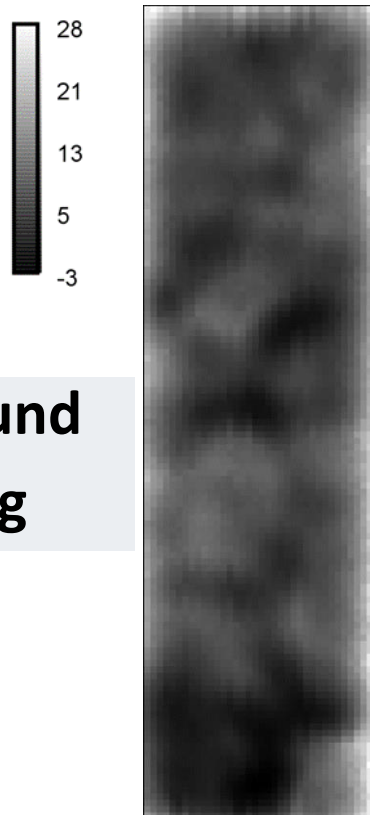
- **Key points to do in the future:**
  - **Measurement in UT of the volume porosity.**
  - **Measurement in UT of the area porosity along the sample thickness.**



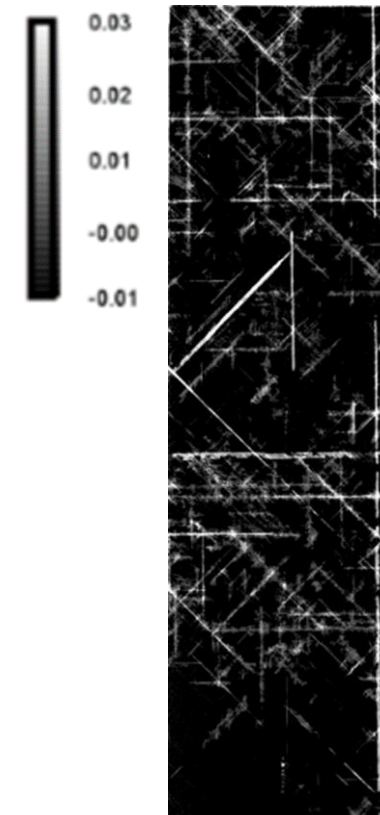
# XCT vs UT

- For a good dataset, dozens of samples must be measured using both techniques.
- The regression-CNN we are using has 6000 points, and the feature extraction is performed in UT and XCT volumes.
- The trained model will learn to improve the measurement performed in UT giving a “XCT-quality” scan.

Ultrasound  
Testing



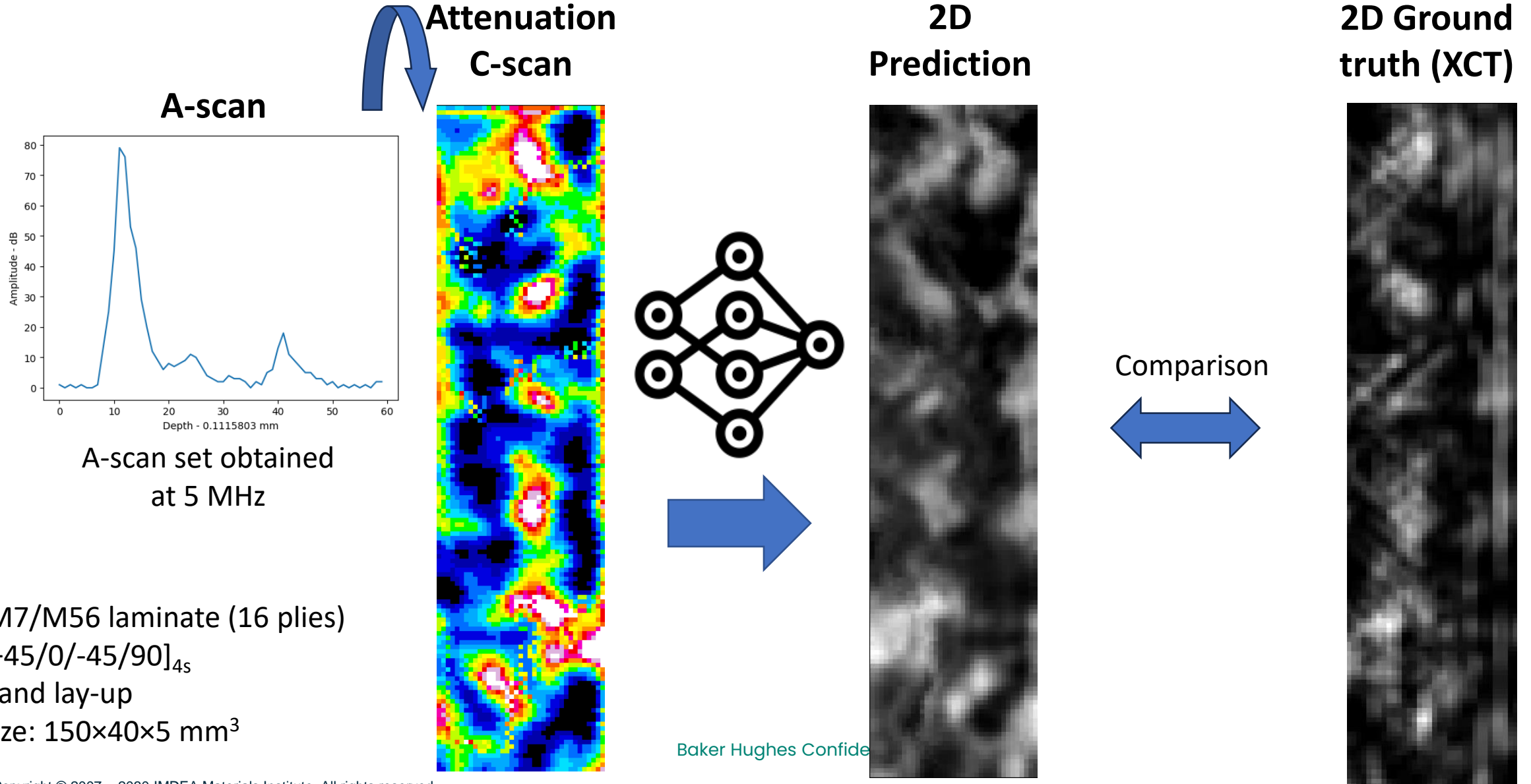
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X-Ray  
Computed  
Tomography



# XCT vs UT



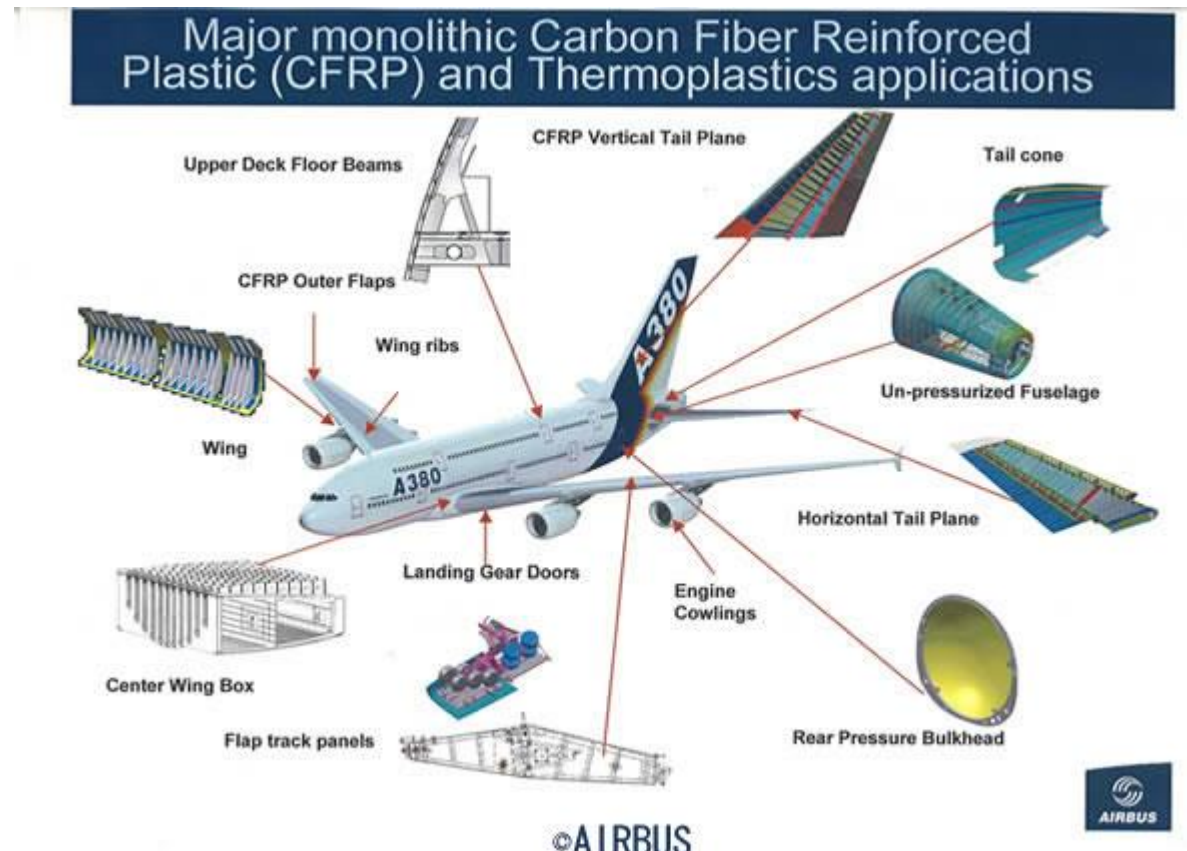
A-scan set obtained  
at 5 MHz

IM7/M56 laminate (16 plies)  
[+45/0/-45/90]<sub>4s</sub>  
Hand lay-up  
Size: 150×40×5 mm<sup>3</sup>

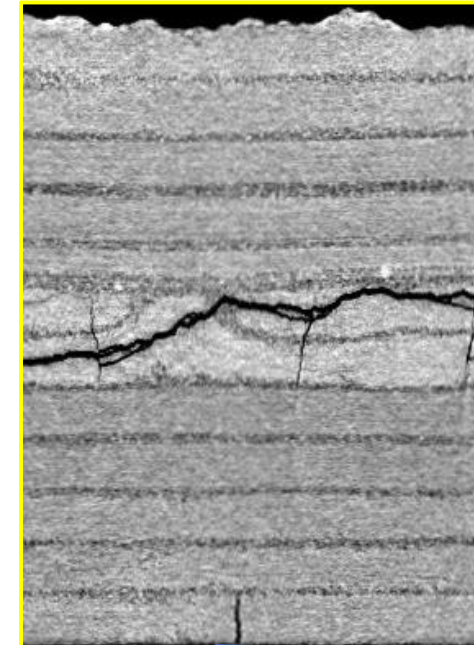
# Damage assessment in CFRP composites after tensile and fatigue tests

# Damage assessment

- Aircraft weight is reduced using CFRP in several parts.
- Less weight means less fuel: cost savings and less pollution.
- However, it is necessary to know the behaviour of CFRP under aggressive conditions.
- Depending on the stacking sequence, cracks and delamination have different development.

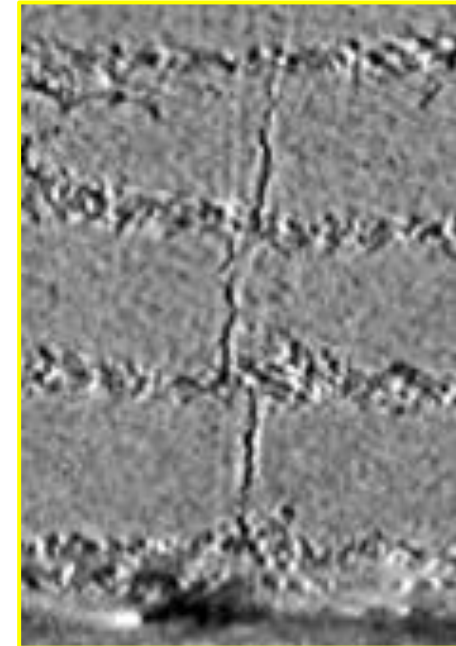


- **Three segmentation paths can be followed:**
  - **BINARIZATION**
    - **Time saving**
    - **Easy to implement**
    - **Less human error**
    - **Implemented in ImageJ**



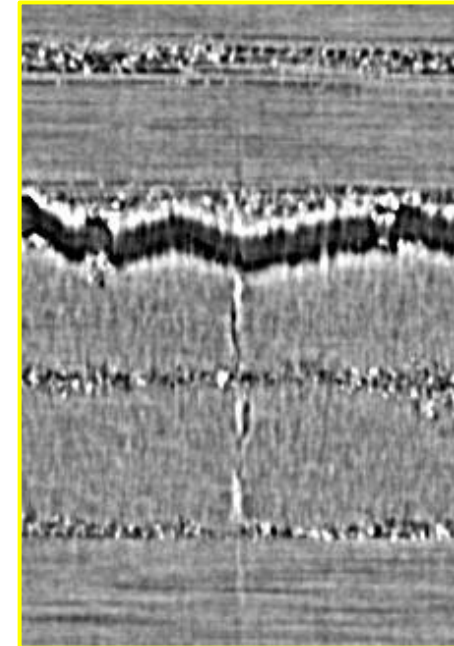
Good contrast, good SNR,  
easy to identify borders  
and defects

- **Three segmentation paths can be followed:**
  - **DEEP LEARNING**
    - Takes time to train the model
    - Easy to implement once the model is trained
    - Implemented in Avizo



Good quality, easy to manually label the defects

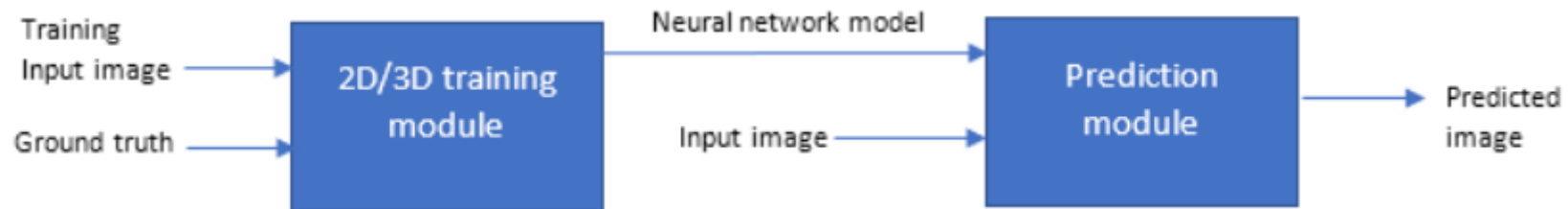
- **Three segmentation paths can be followed:**
  - **MANUAL**
    - **Time consuming**
    - **Non generalizable for other samples**
    - **High human error**
    - **Implemented in ImageJ and Avizo**



High phase contrast  
enhancement

# Damage assessment

- Avizo implements several modules focused on deep learning methods.
- Two phases:
  - Training
  - Prediction

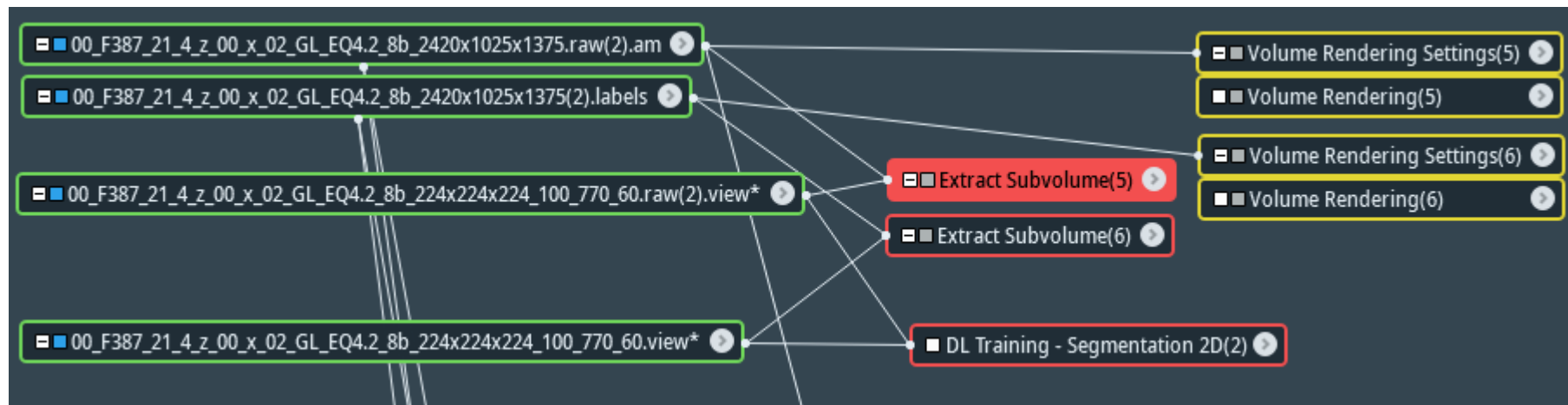


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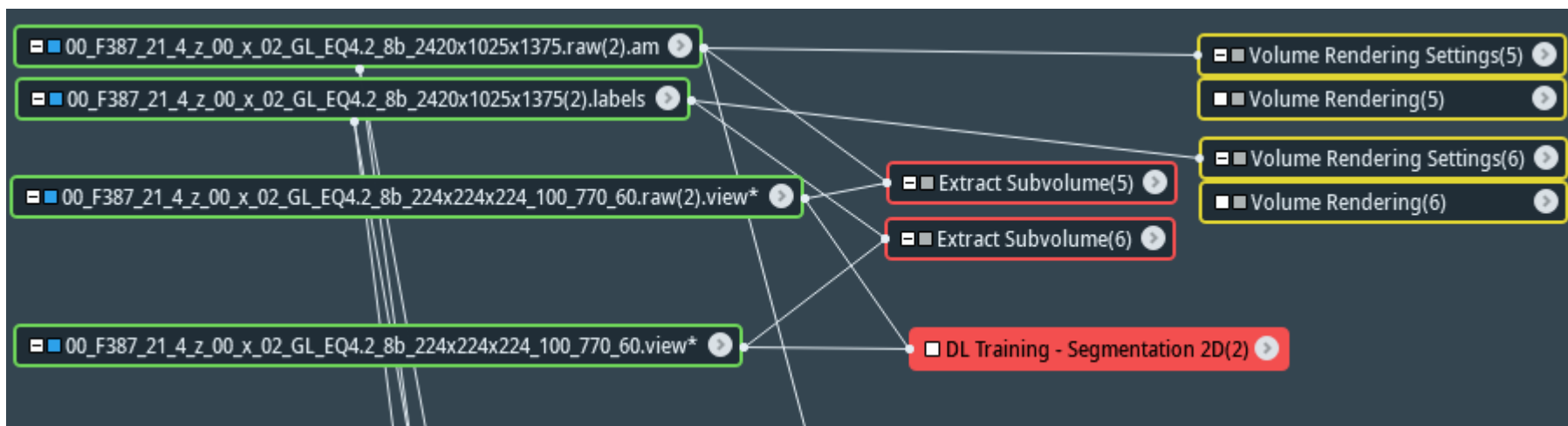
- **TRAINING:**
- **Segmented ground truth is necessary. Labelling will help to classify different defects (AI Assisted Segmentation, implemented in Avizo, too).**
- **The module needs as inputs the segmented ground truth and the grey-level original.**
- **Avizo uses a shallow fully convolutional neural network for image semantic segmentation.**
- **Possibility to select the number of epochs and elastic deformation.**
- **The training can be done at specific regions, but the prediction is computed on the whole volume.**



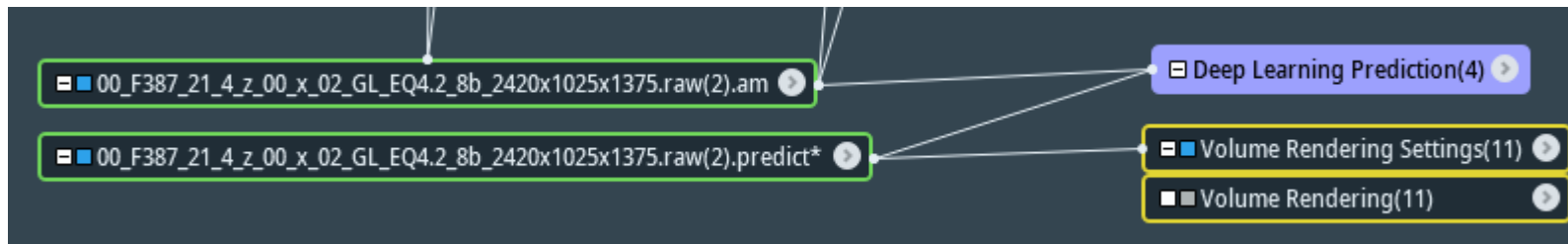
- **TRAINING:**
- **Extract Subvolume** to extract regions from the ground truth grey-level volume and its labelled counterpart.
- **Cubic subvolumes** with the feature we need to identify are the best choice.
- **DL Training – Segmentation 2D** module trains the set with variants of U-Net model (VGG, ResNet...).



- **TRAINING:**
- Train Learning Curve and Validation Learning Curve will verify the goodness of the model.
- Loss function must be at the minimum value and accuracy near 1.
- Try to avoid underfitting and overfitting!
- Check the proportion between train and validation datasheets.

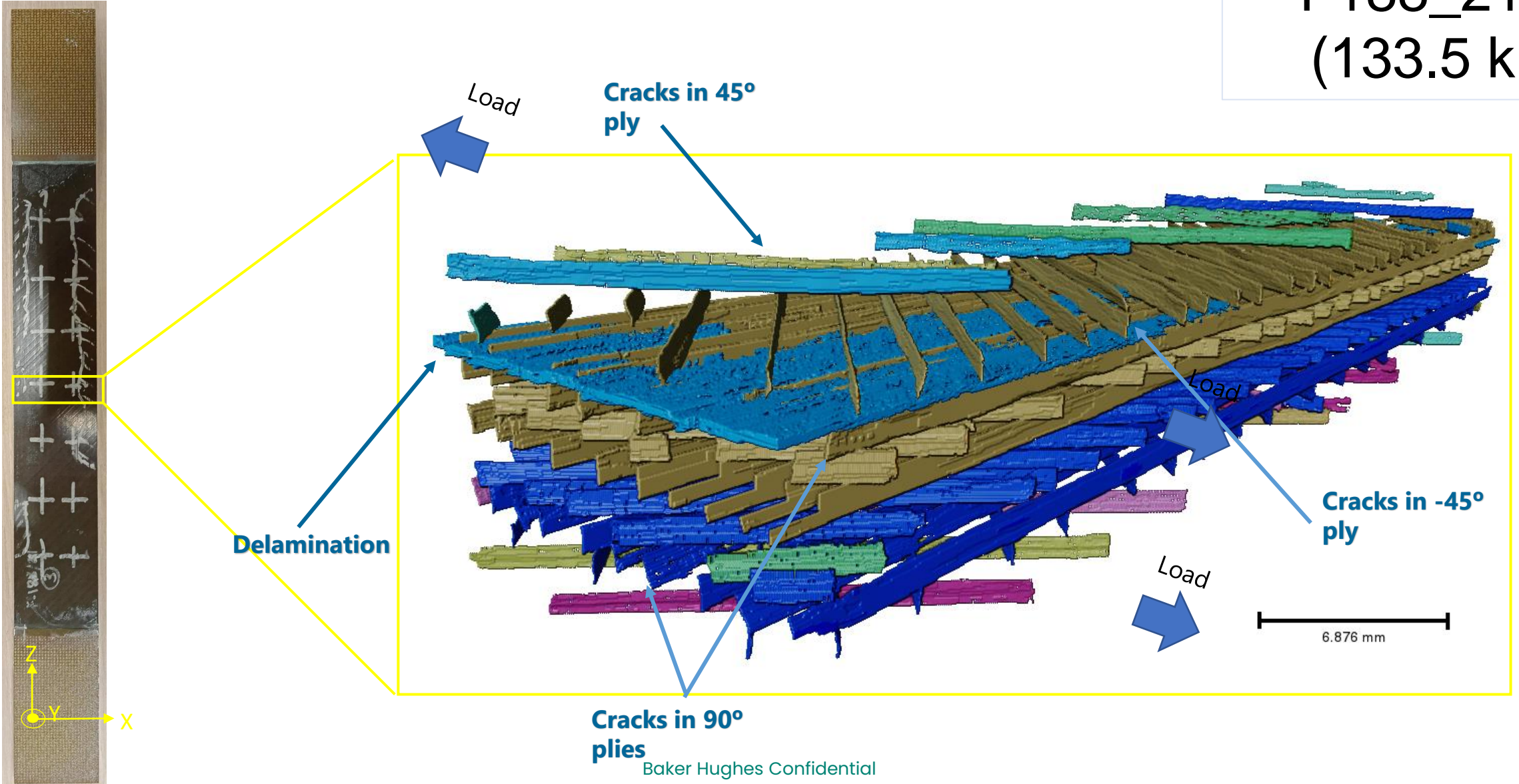


- **PREDICTION:**
- **Load the volume to segment. It has to be similar to the ground truth volume.**
- **The trained classifier has the architecture and weights. These parameters can be adjusted, too.**
- **The prediction can be done for the whole volume or applied as overlapped regions.**



# Damage assessment

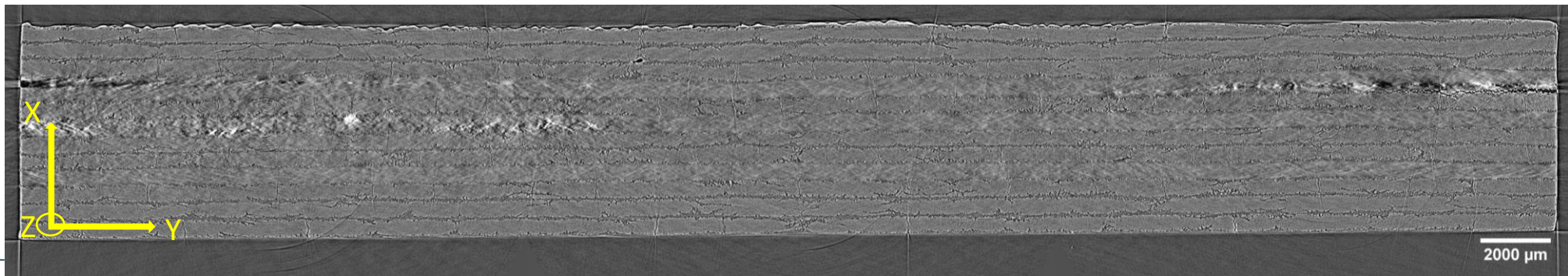
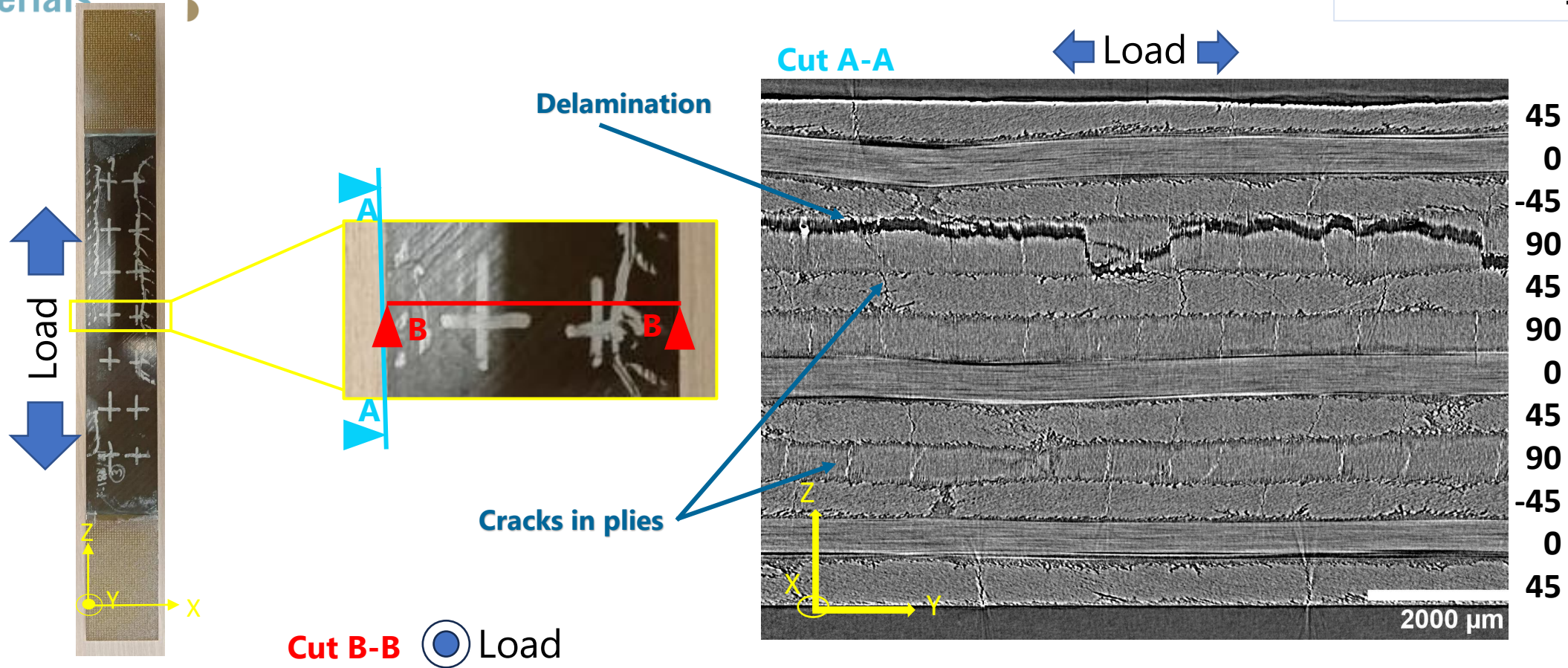
F188\_21\_3  
(133.5 kN)



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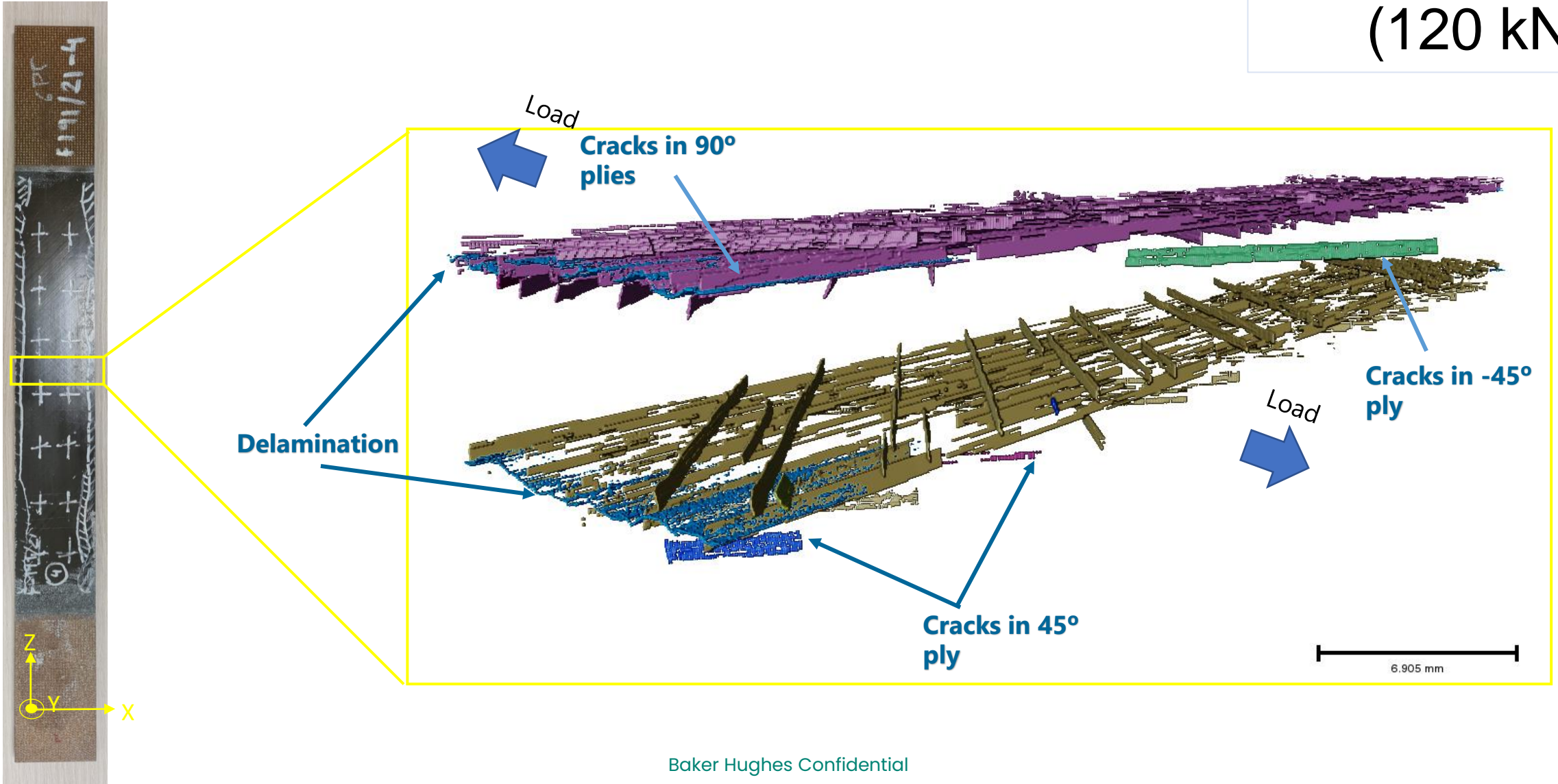
# Damage assessment

F188\_21\_3



# Damage assessment

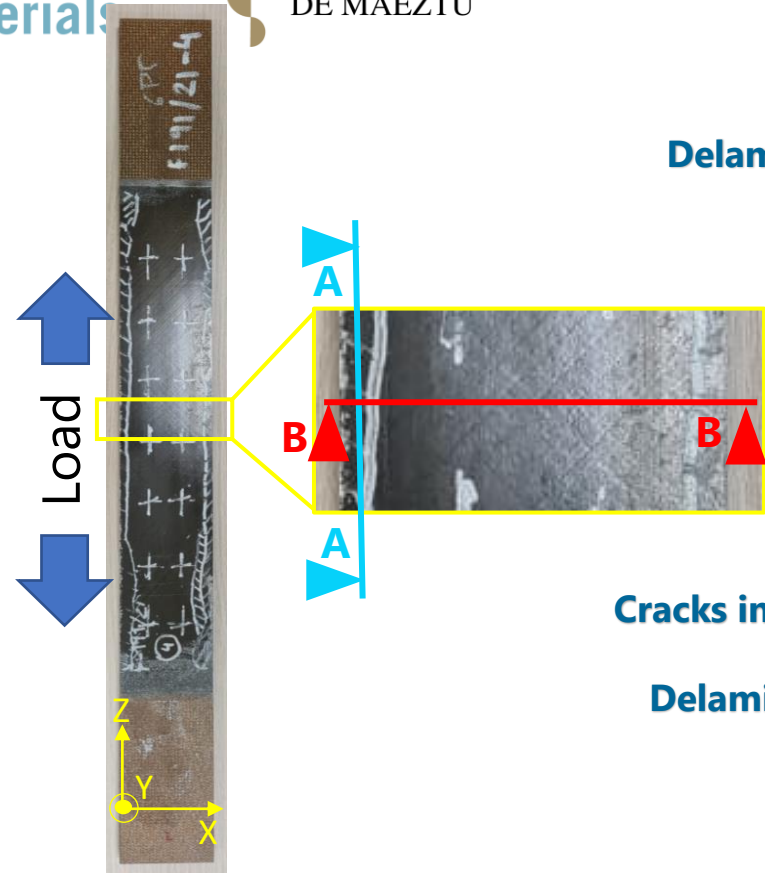
F191\_21\_4  
(120 kN)



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# Damage assessment

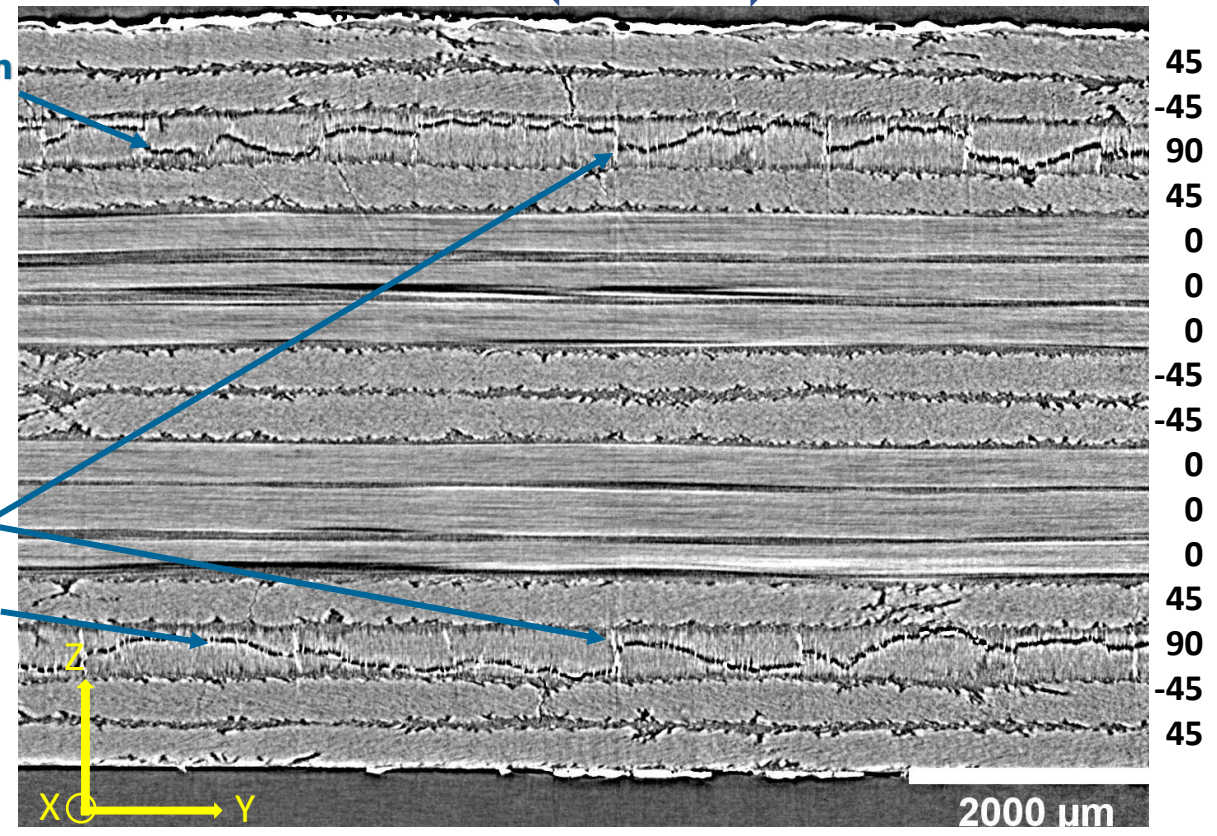
F191\_21\_4



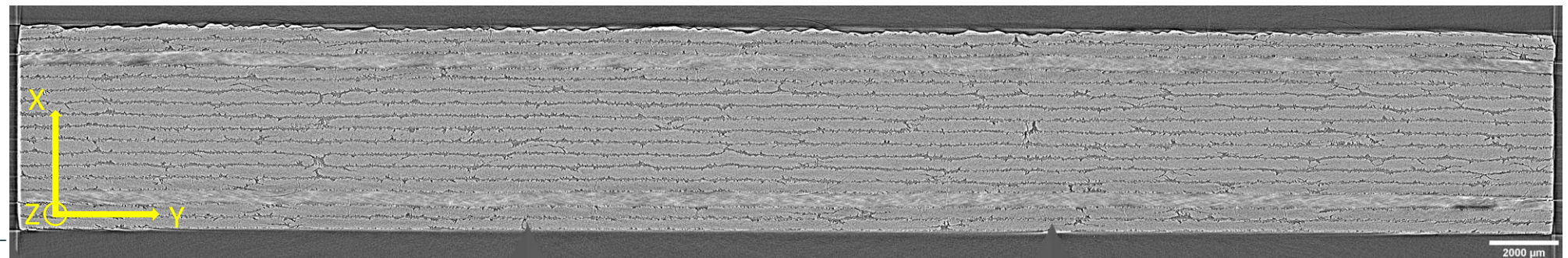
Delamination

Cut A-A

← Load →

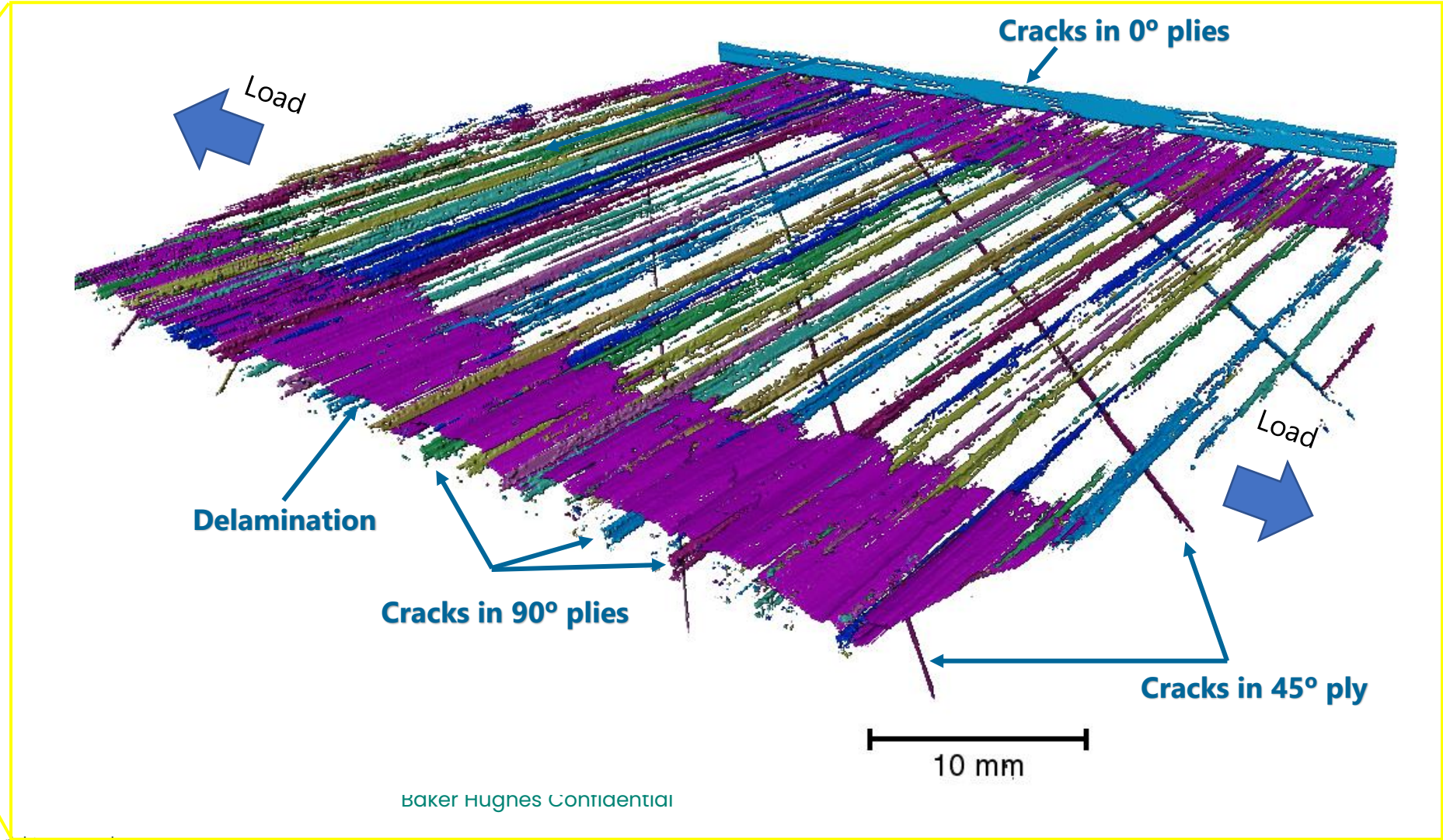


Cut B-B  Load



# Damage assessment

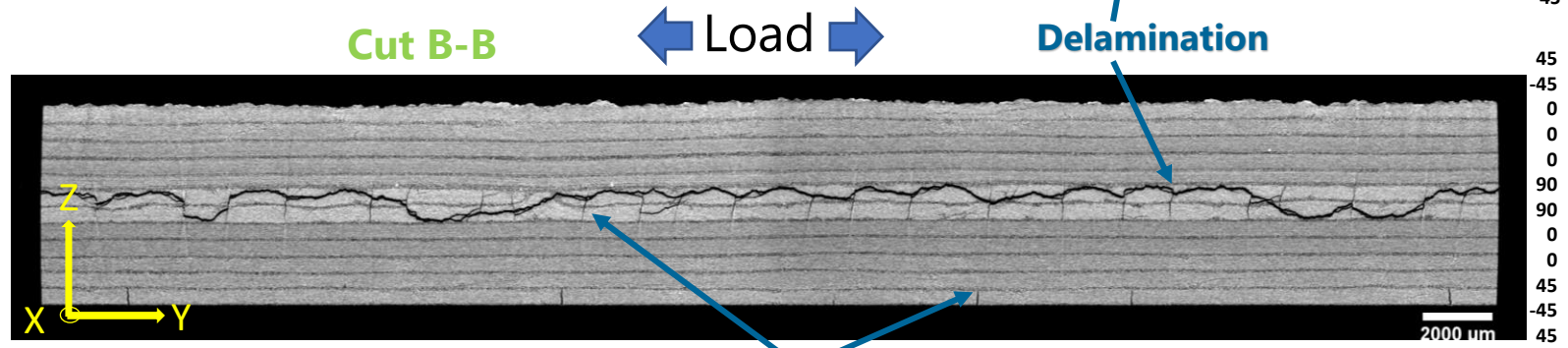
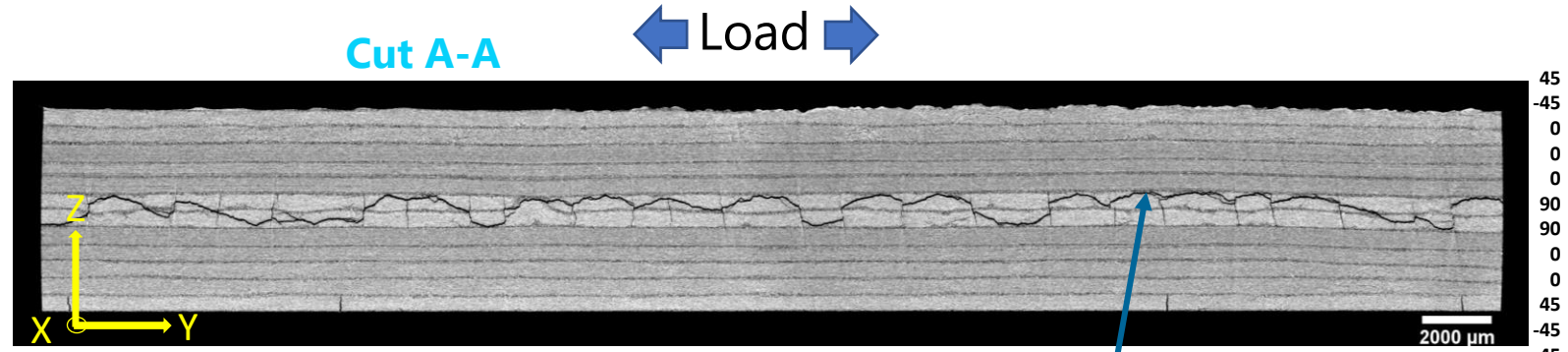
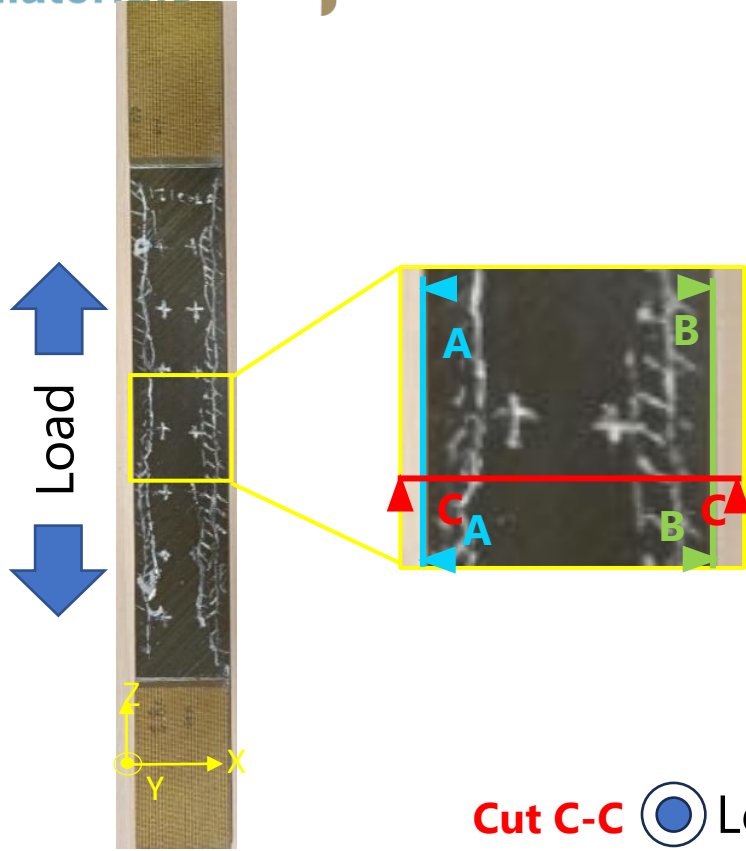
F387\_21\_1  
(90 kN)





# Damage assessment

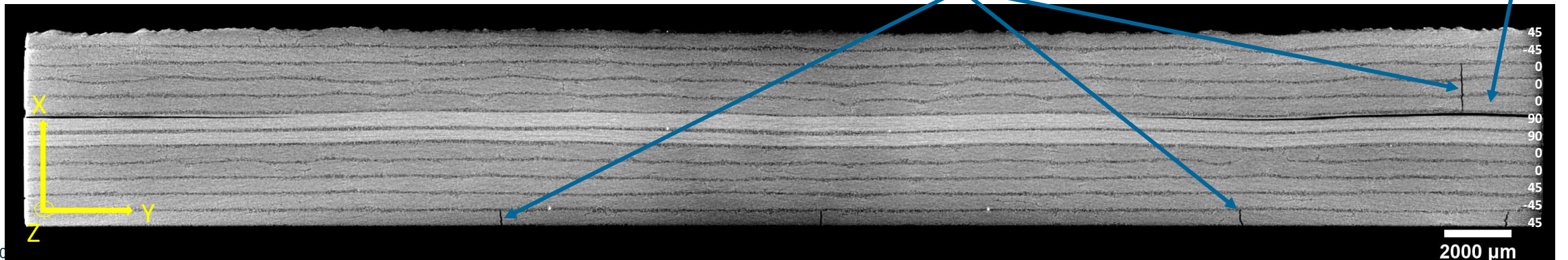
F387\_21\_1



Cut C-C  Load

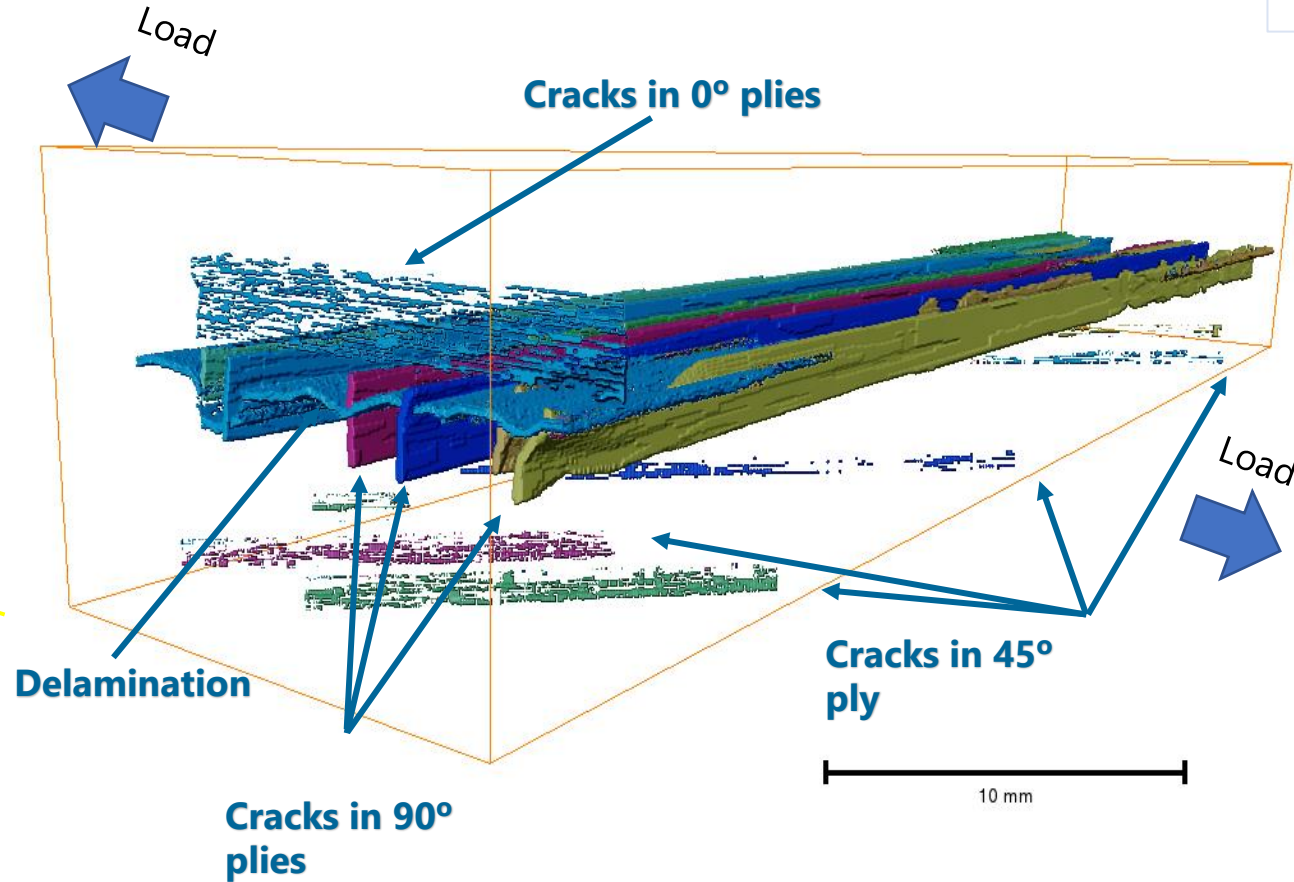
Cracks in plies

Delamination



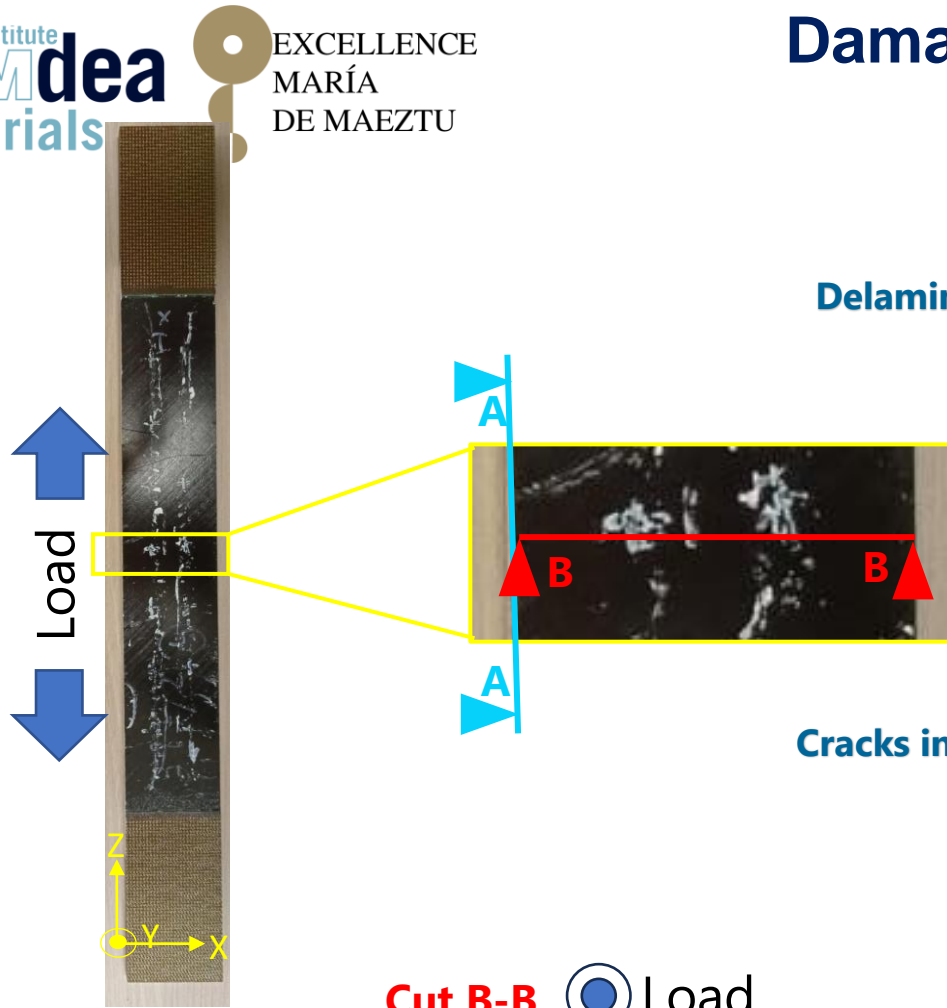
# Damage assessment

F387\_21\_4  
(120 kN)

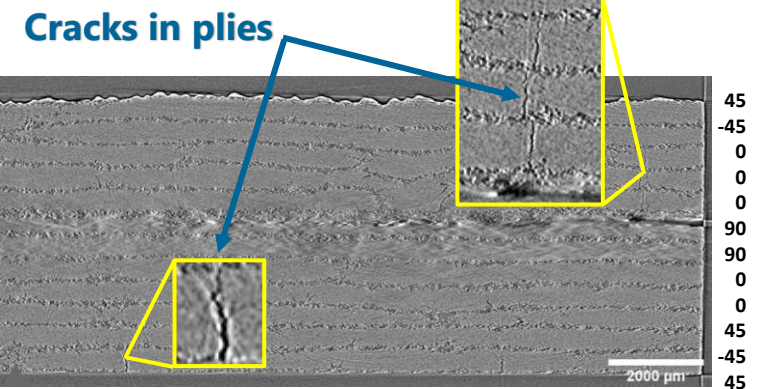
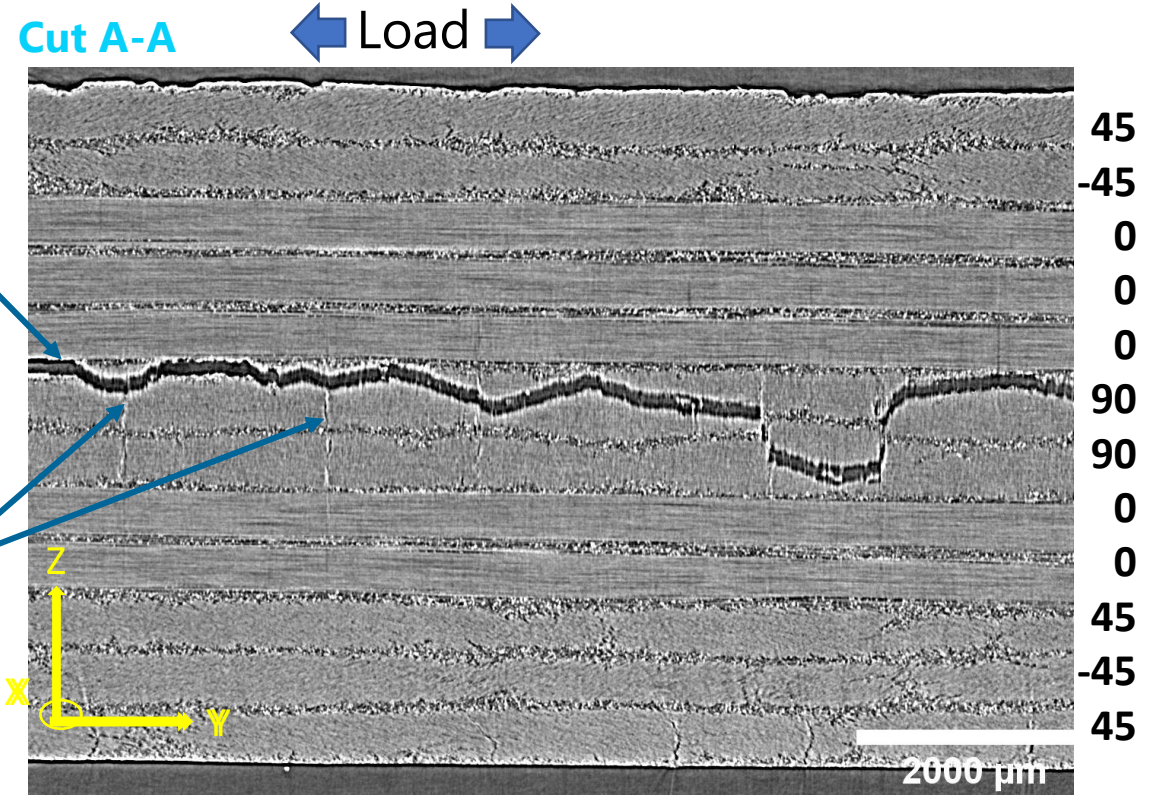


# Damage assessment

F387\_21\_4

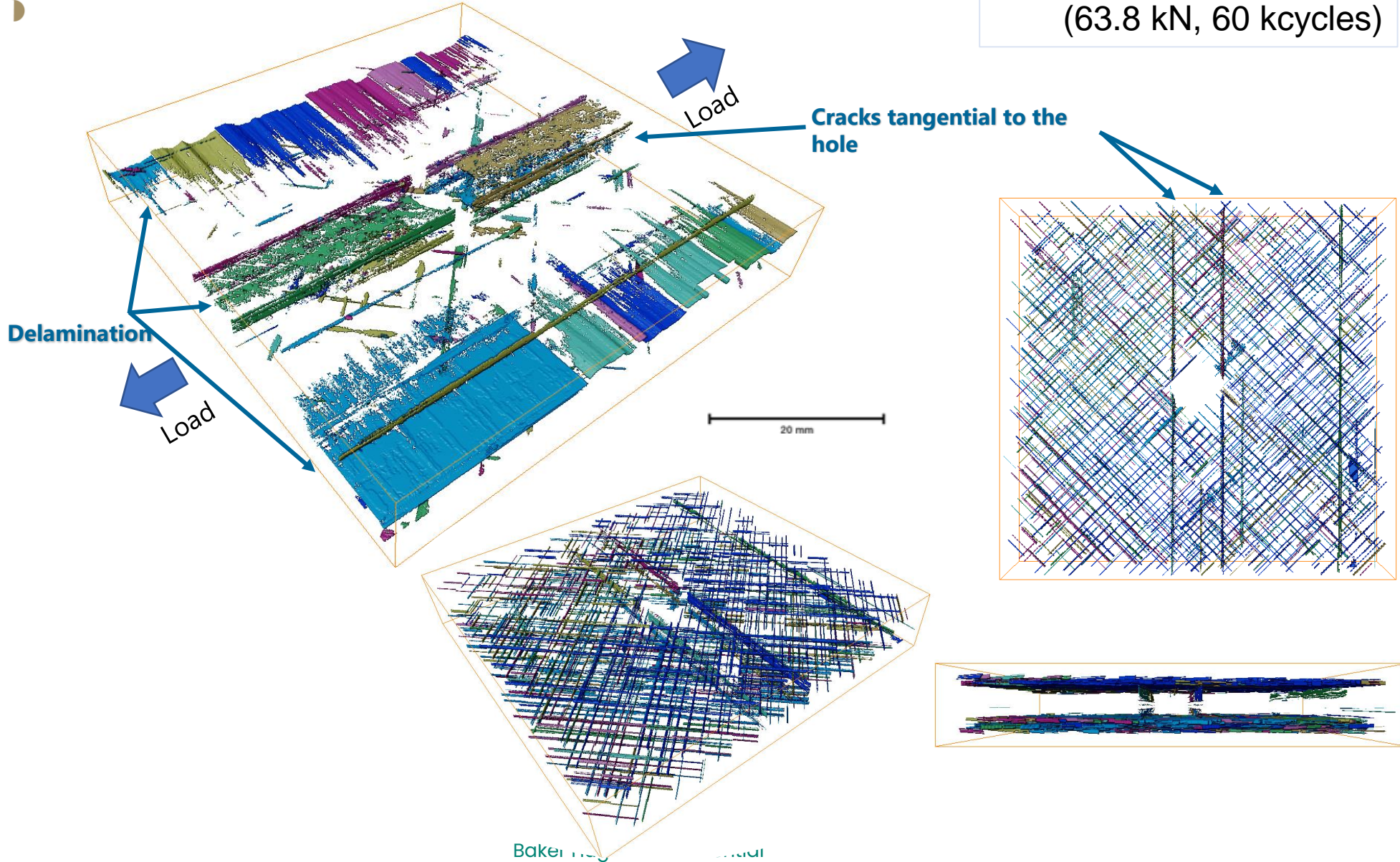


Cut B-B  Load



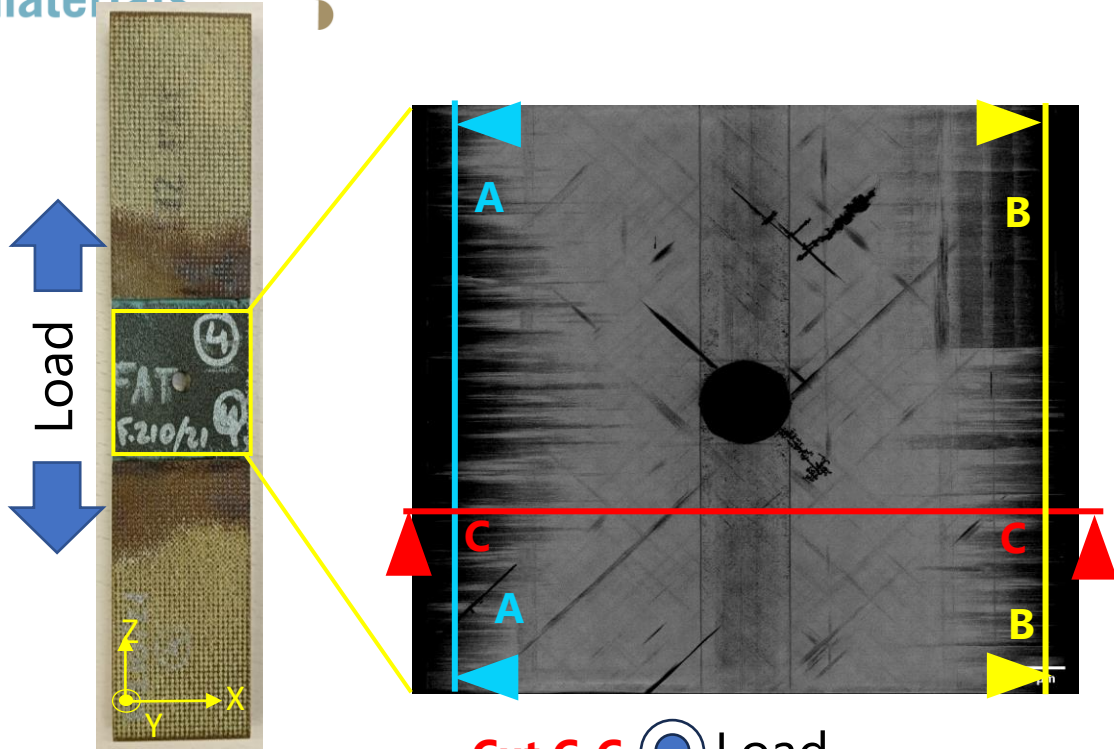
# Damage assessment

F210\_21\_4  
(63.8 kN, 60 kcycles)



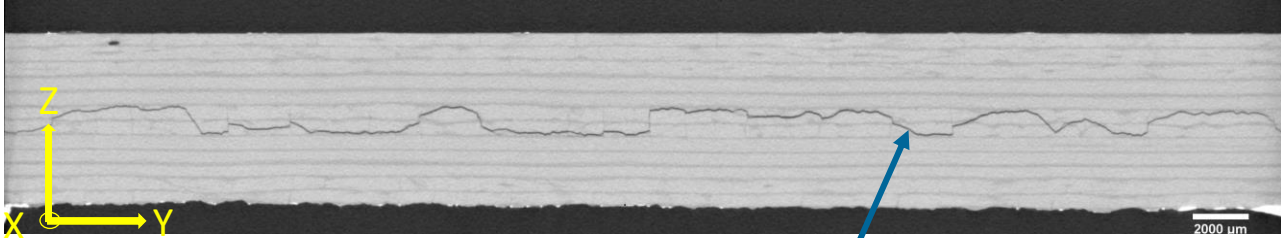
# Damage assessment

F210\_21\_4

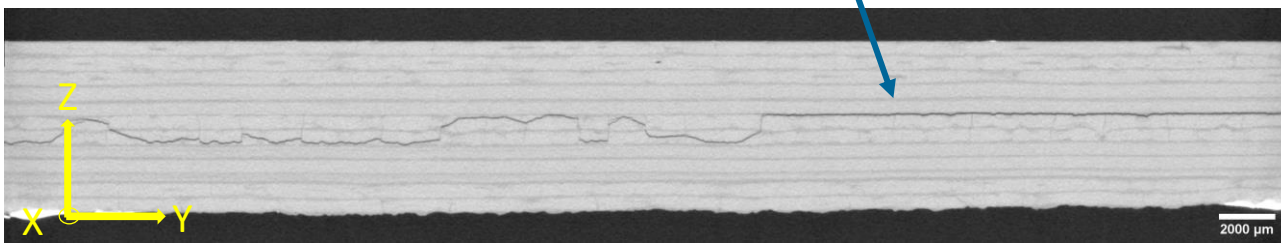


← Load →

Cut A-A



Cut B-B

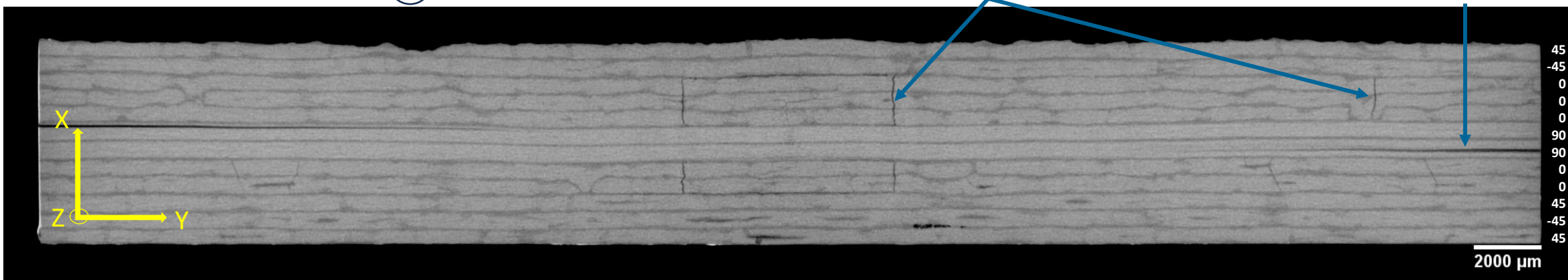


Delamination

Cut C-C  Load

Cracks in plies

Delamination

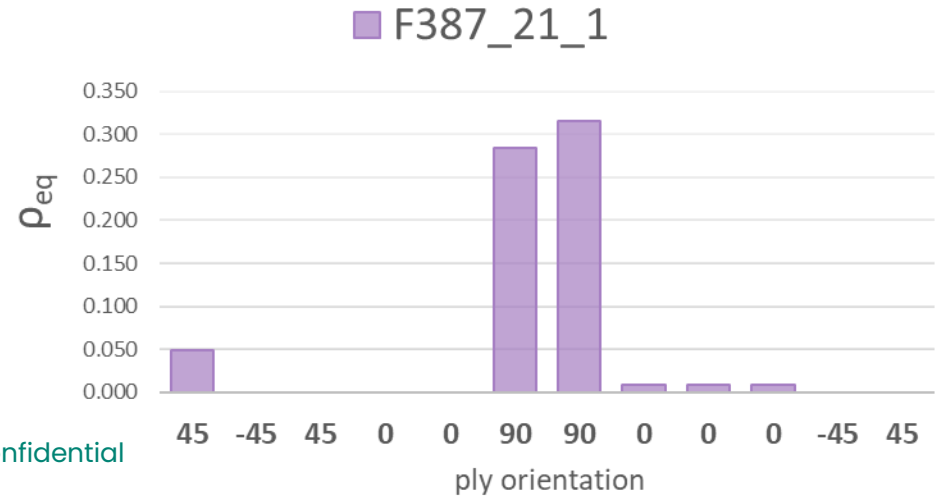
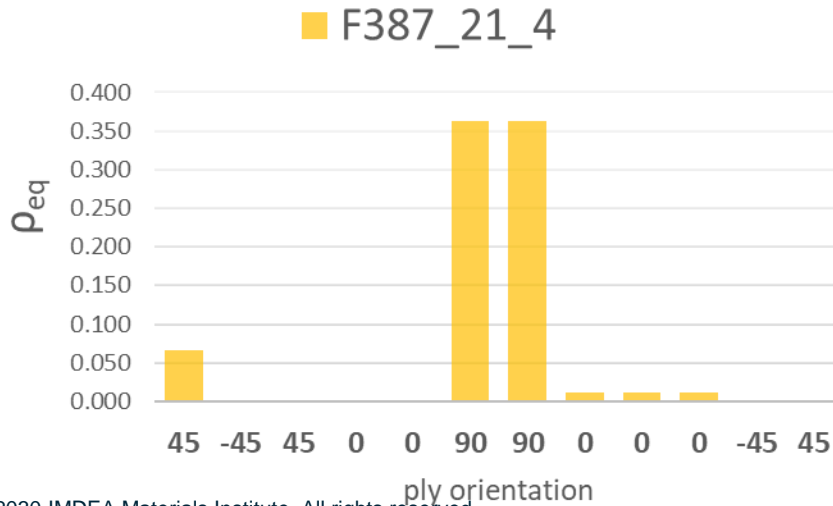
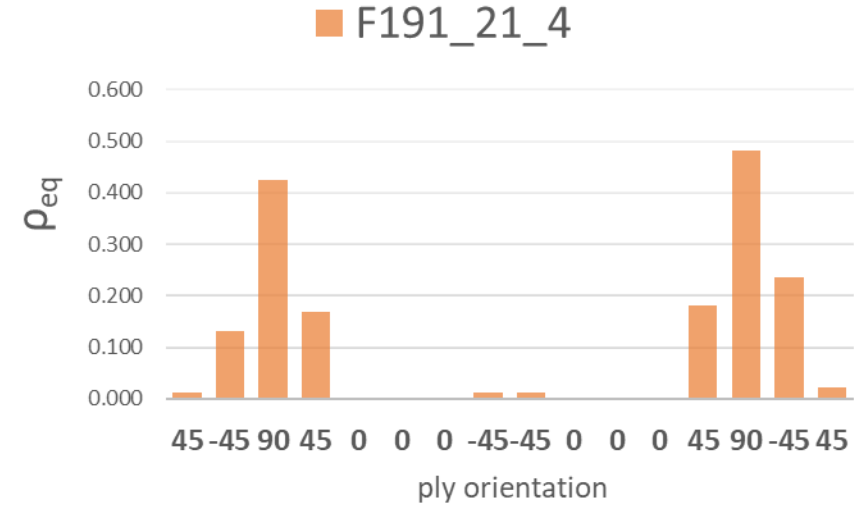
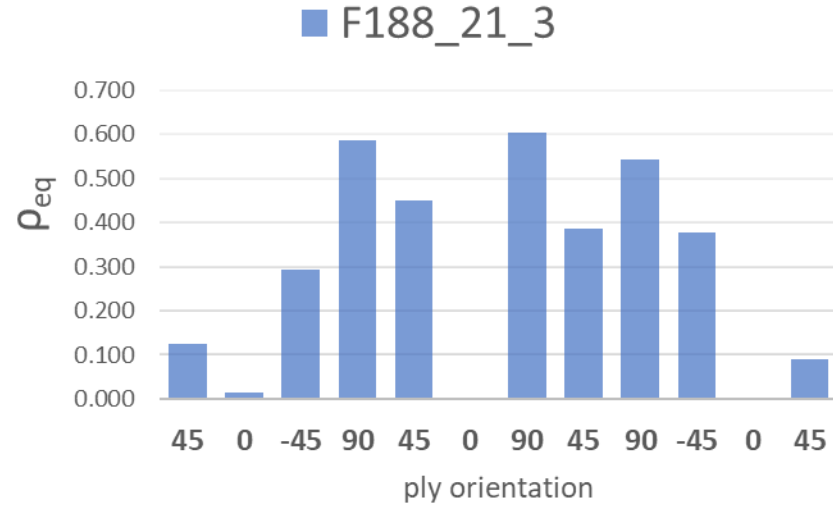


# Damage assessment

$$\rho_{eq} = \frac{L \cdot t}{A}$$

- $\rho_{eq}$  = Crack density per ply
- $L$  = Sum of the length of each crack in the ply
- $t$  = Thickness of the ply
- $A$  = Area of the analyzed ply

## Crack density per ply in tensile samples

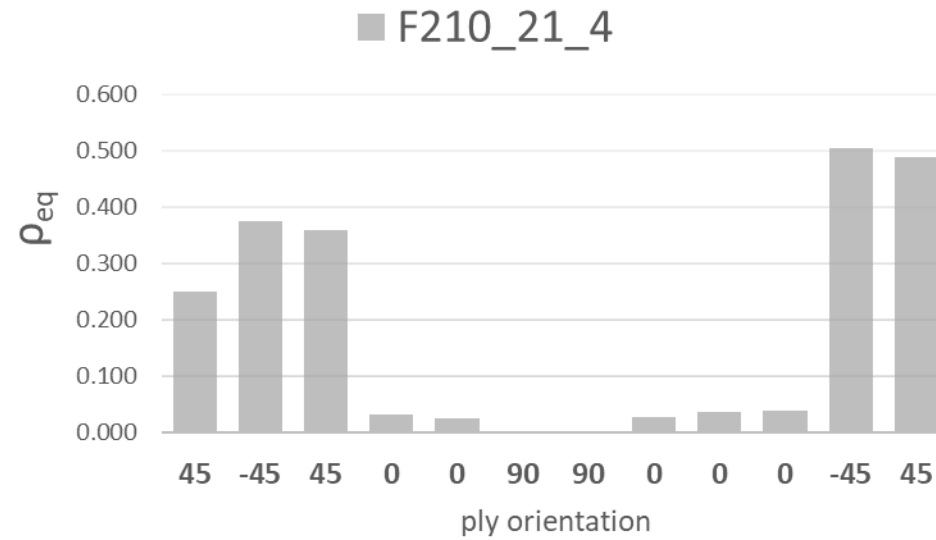


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# Damage assessment

$$\rho_{eq} = \frac{L \cdot t}{A} \begin{cases} \rho_{eq} = \text{Crack density per ply} \\ L = \text{Sum of the length of each crack in the ply} \\ t = \text{Thickness of the ply} \\ A = \text{Area of the analyzed ply} \end{cases}$$

Crack density per ply in fatigue sample

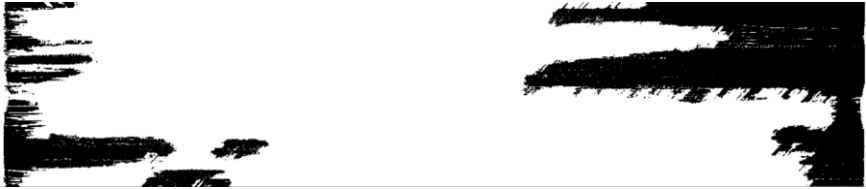


# Damage assessment

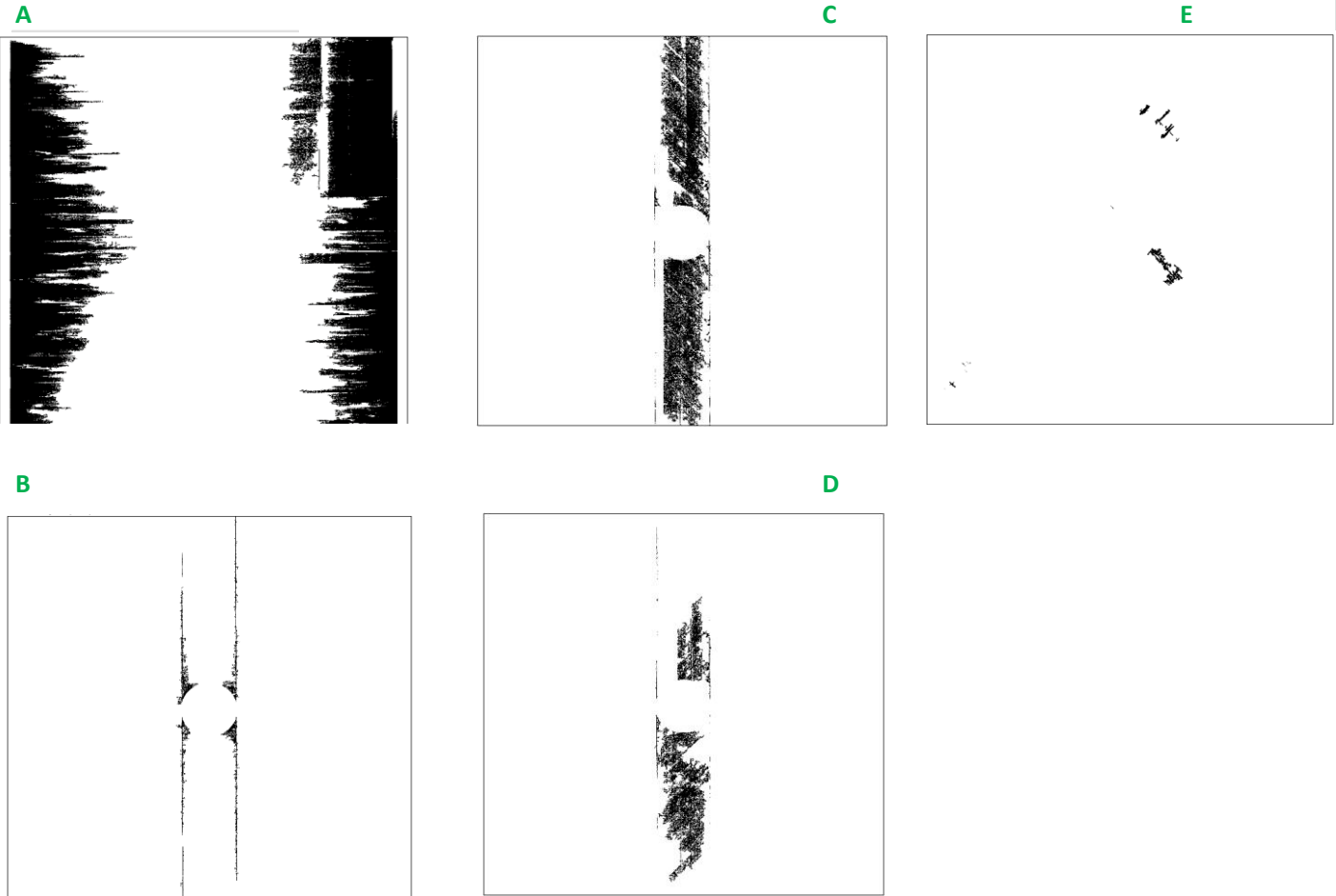
$$A_{\text{delaminated}}/A_{\text{analyzed}} = \frac{\text{Area}}{\text{Nominal Area}}$$

## Delaminated area in all samples

F188\_21\_3



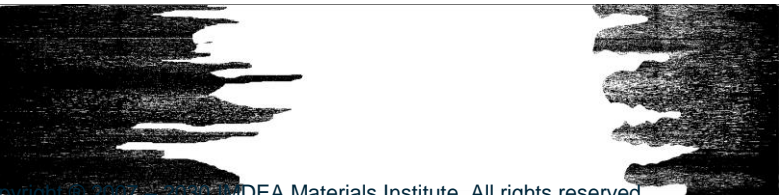
F210\_21\_4



F191\_21\_4



F387\_21\_4



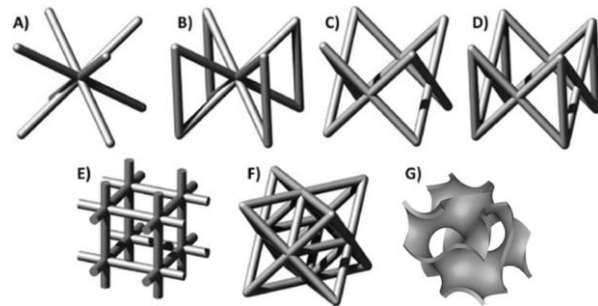


# Understanding degradation behavior of Mg scaffolds manufactured via LPBF

- **Materials in health must fulfil three requirements:**
  - **Biocompatibility**
  - **Biodegradation**
  - **Similar mechanical properties**

# Degradation of scaffolds

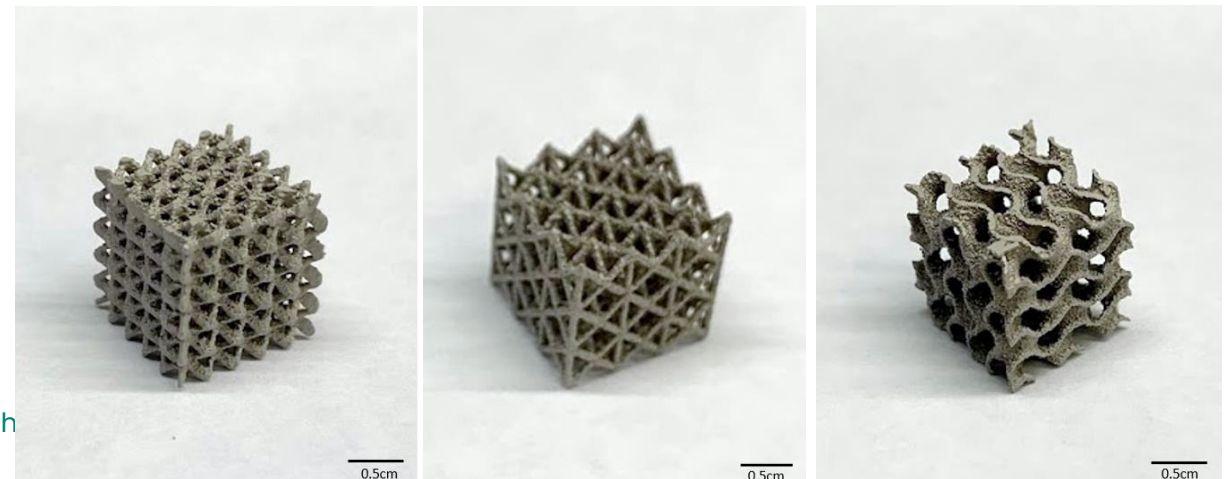
- **Magnesium alloys can verify these requirements.**
  - **Avoids stress shielding and body rejection.**
  - **Rapid degradation rate.**
- **Additive Manufacturing (AM) can build complex geometries that can be fitted to precise body parts. Selective Laser Melting (SLM) is a good choice.**
  - **High precision and flexibility fabrication**
  - **Enhancement of surface properties after Plasma Electrolytic oxidation (PEO) treatment.**



BCC

FCC

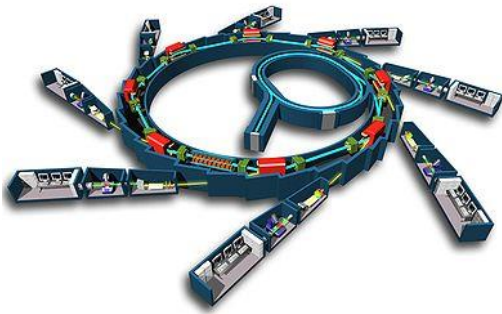
GYROID



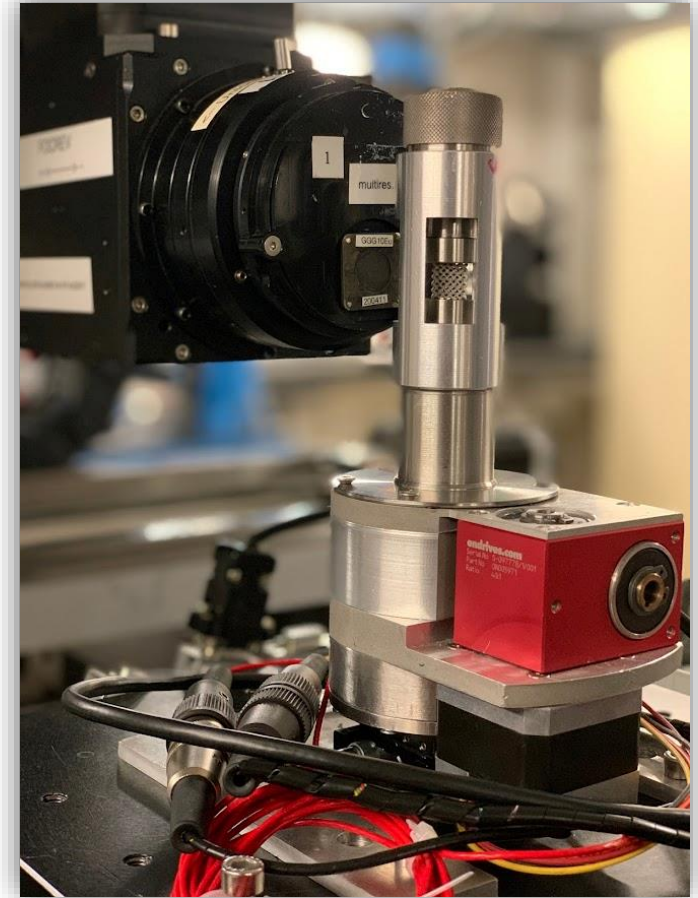
Baker Hugh

# Degradation of scaffolds

- WE43-PEO BCC, FCC, and gyroid scaffolds were built.
- They were immersed into SBF for 0, 7, 14, and 21 days.
- Interrupted in-situ compression tests were conducted at room temperature at the microtomography ID19 beamline of the European Synchrotron Radiation Facility (ESRF) in Grenoble (France).
  - Voxel size: 6.5  $\mu\text{m}$ .
  - Monochromatic energy: 80 keV.
- In situ test machine developed at IMDEA Materials.

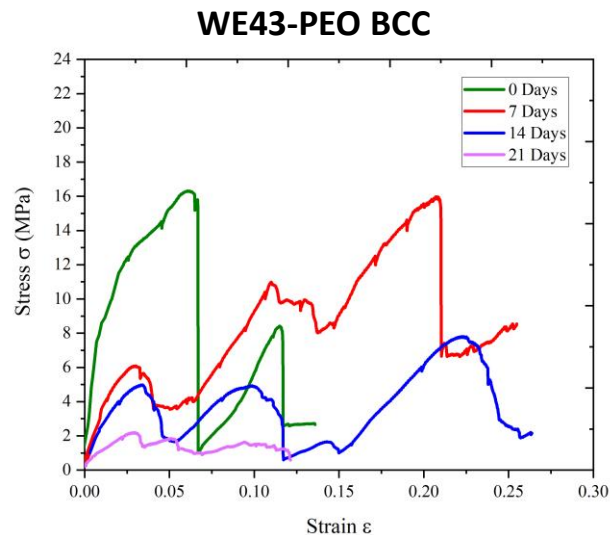


The European Synchrotron **Baker Hughes Confidential**

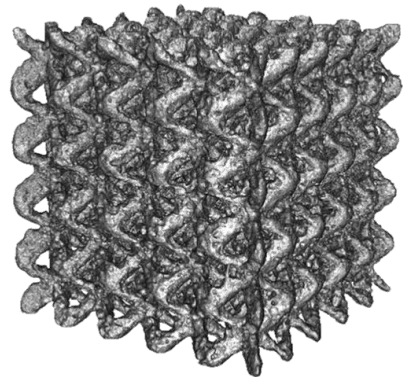


# Degradation of scaffolds

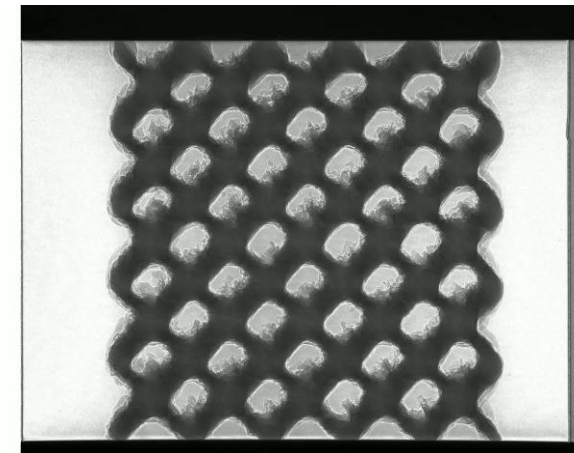
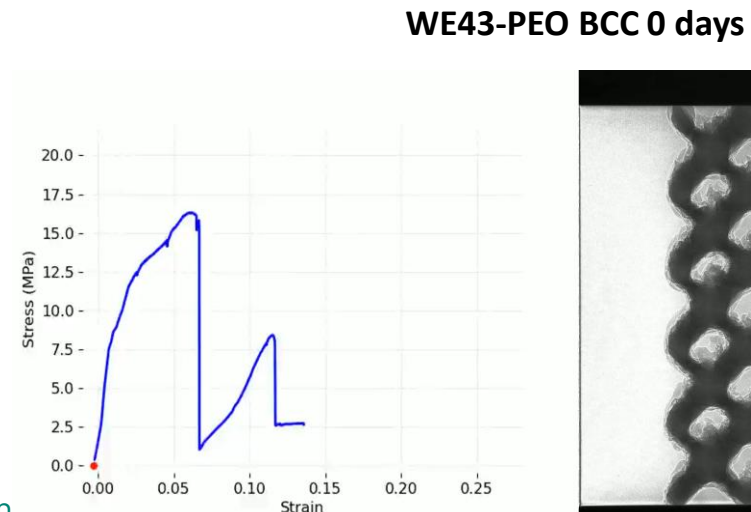
- The maximum strength of the scaffolds decreases substantially when immersing in SBF.
- Only BCC scaffold will be shown in this presentation.
- Digital Image Correlation (DIC) and Digital Volume Correlation (DVC) are implemented in all the samples to evaluate the damage evolution.



**WE43-PEO BCC 0 days**

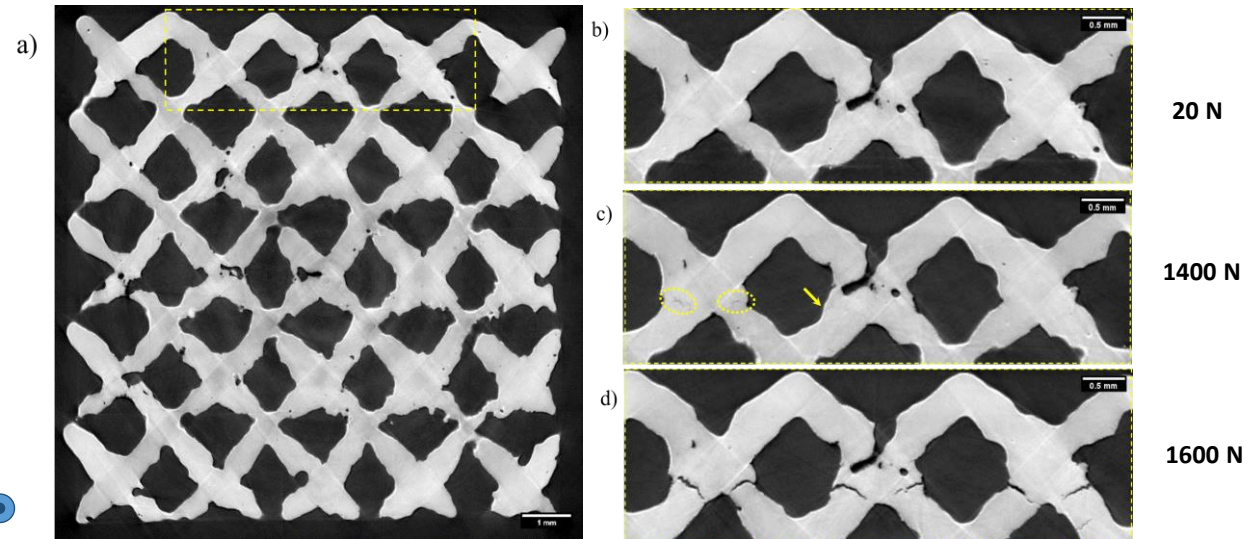
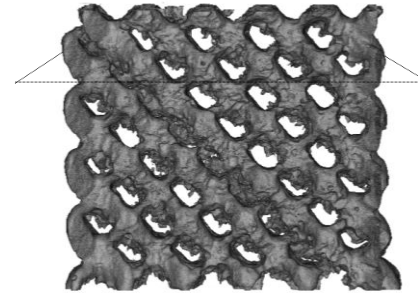
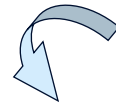


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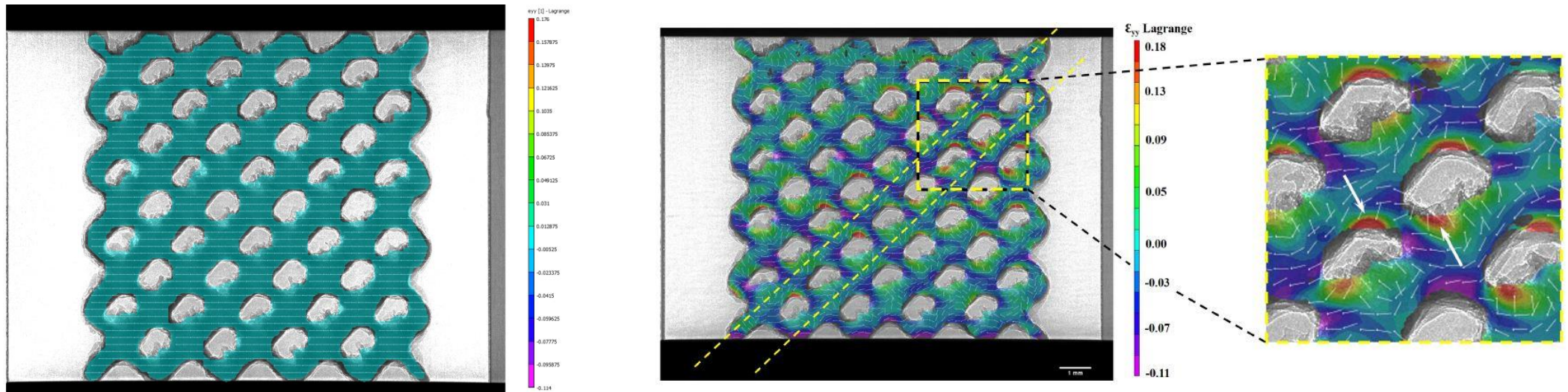
# Degradation of scaffolds

WE43-PEO BCC

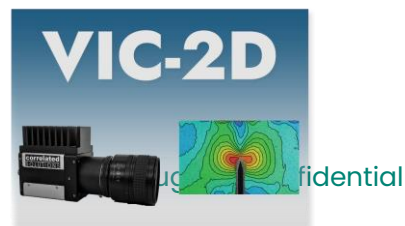


# Degradation of scaffolds

- BCC structure was found to be liable for development of localized deformations.
- Global strain:  $\epsilon = 11\%$ .



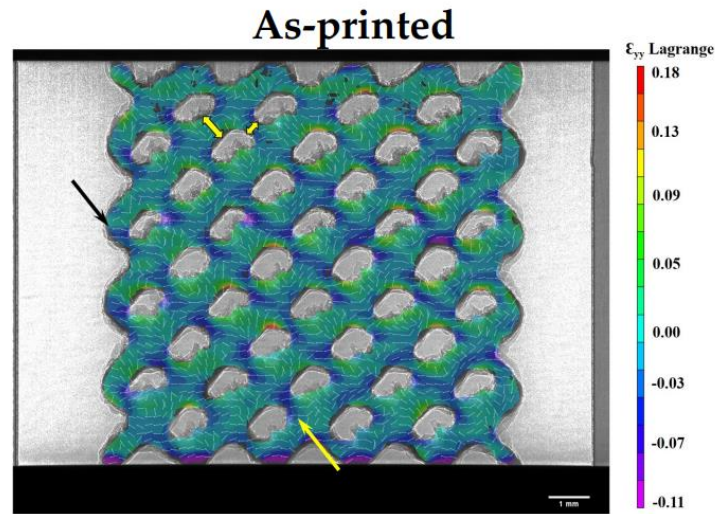
(a) BCC scaffold  $\epsilon=11\%$



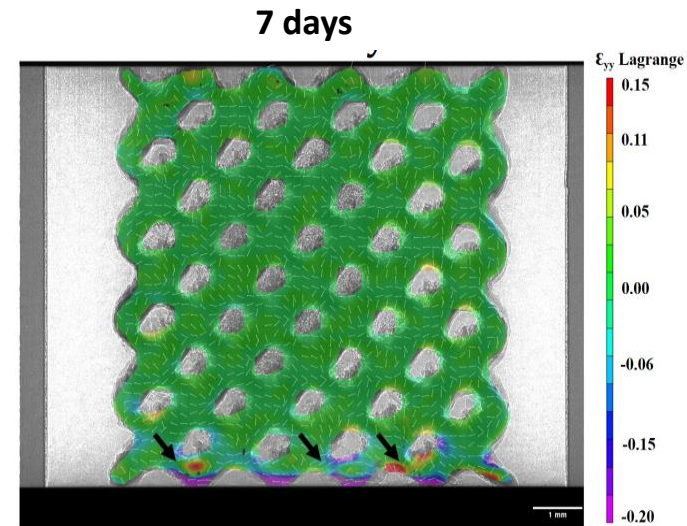
# Degradation of scaffolds

- **AM defects, like heterogeneous strut size, serves as stress concentrators for crack initiation.**
- **Degraded parts after SBF immersion caused failure location.**

Blue deformations are more prevalent in some diagonal struts, indicating higher compressive deformation.



(a) BCC scaffold  $\epsilon=9\%$



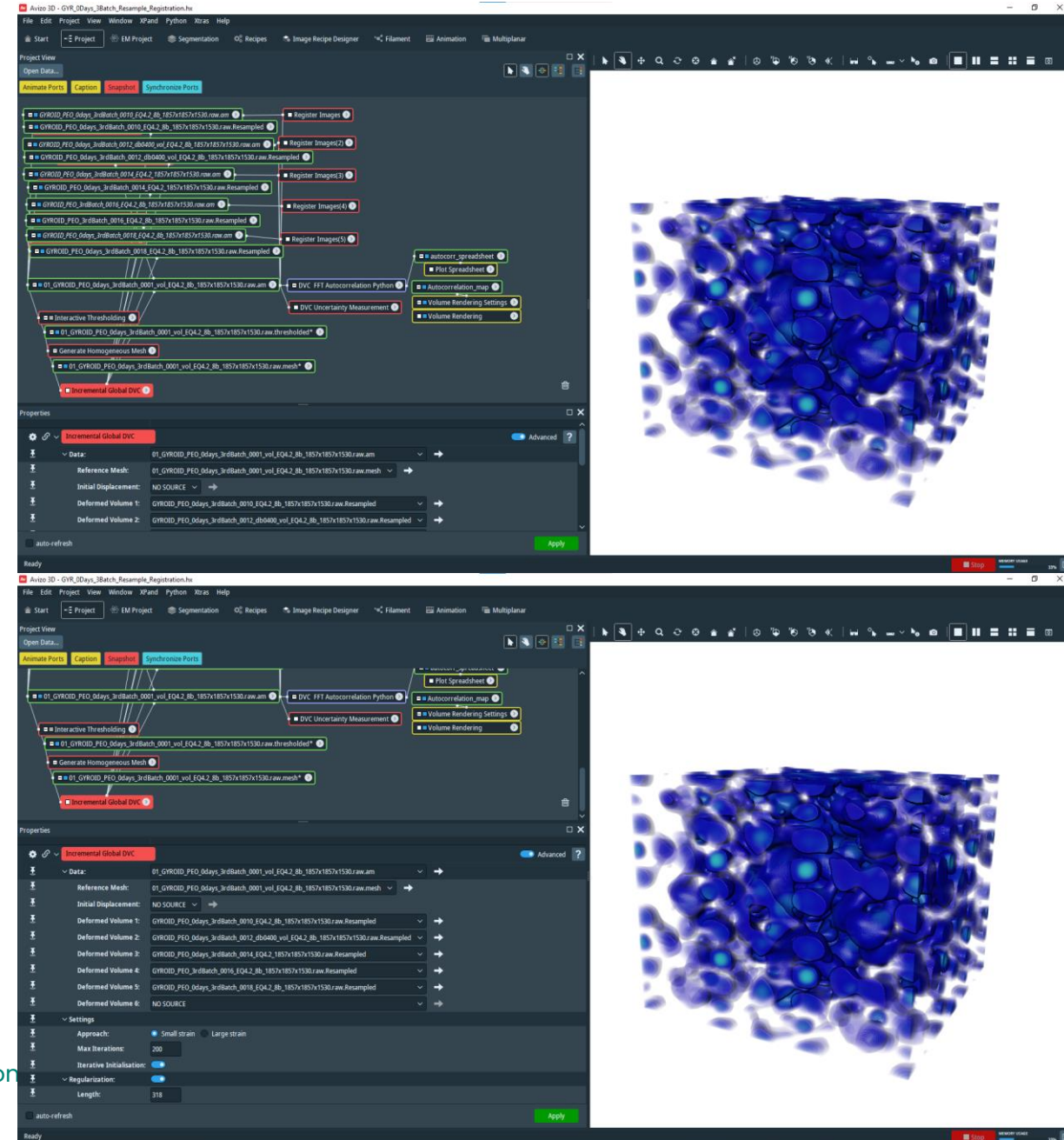
(b) BCC scaffold  $\epsilon = 7\%$

Failure location is transferred to the more damaged parts, not the diagonal struts.



# Degradation of scaffolds

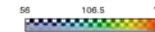
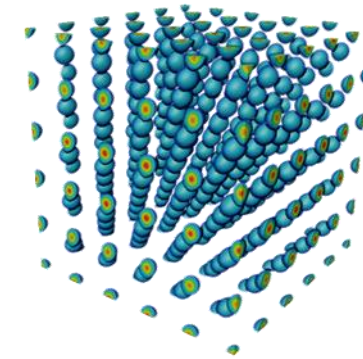
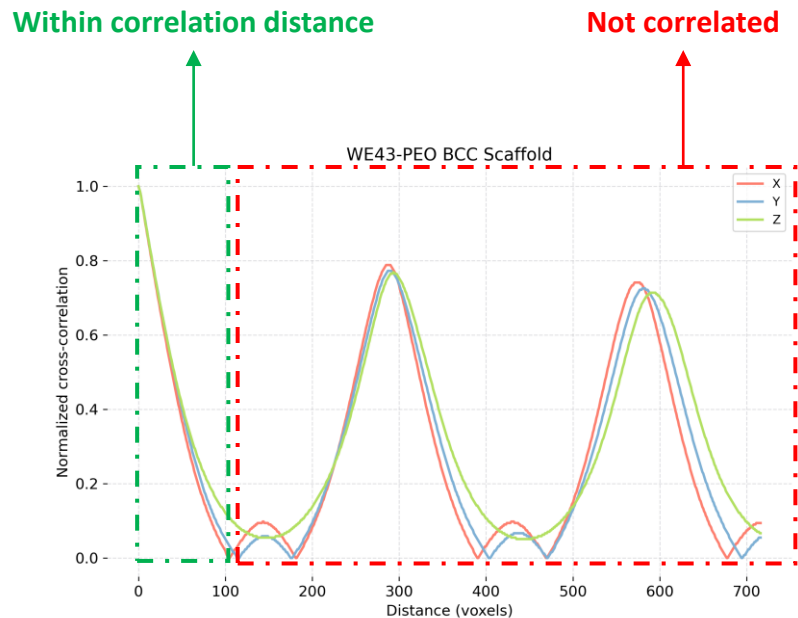
- DVC is an Avizo module:
  - Successive steps are not very distant from each other.
  - Registration is mandatory.
  - Resampling can reduce the computing load.
  - Autocorrelation.
  - Uncertainty.
  - Meshing.
  - Global DVC.



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- Autocorrelation:**

- DVC relies on naturally occurring texture to perform registration between two successive 3D volumes.
- Correlation measurements reveal the spatial arrangement of features within the material's microstructure.
- They guide the selection of an optimal mesh cell size.

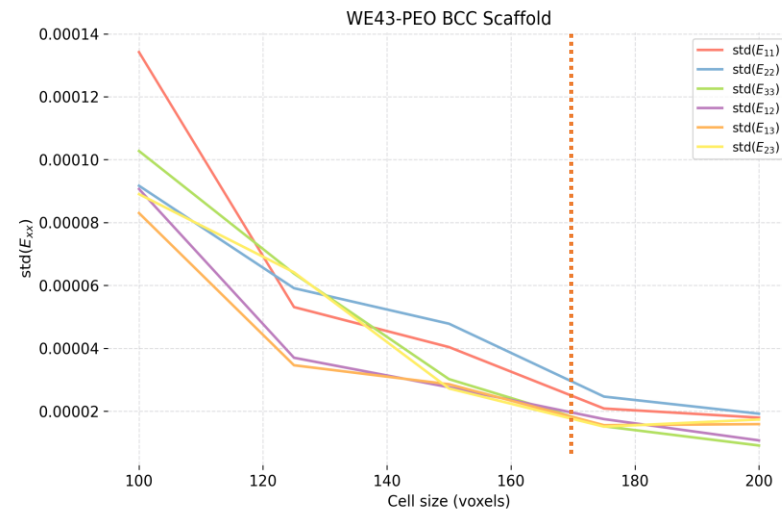


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# Degradation of scaffolds

- **Uncertainty:**

- Once a range of mesh sizes was selected, the uncertainty assessment for different mesh sizes was carried out to find out the strain error for each mesh size.
- Resolution and precision were obtained, with strain errors no greater than 0.08 %.

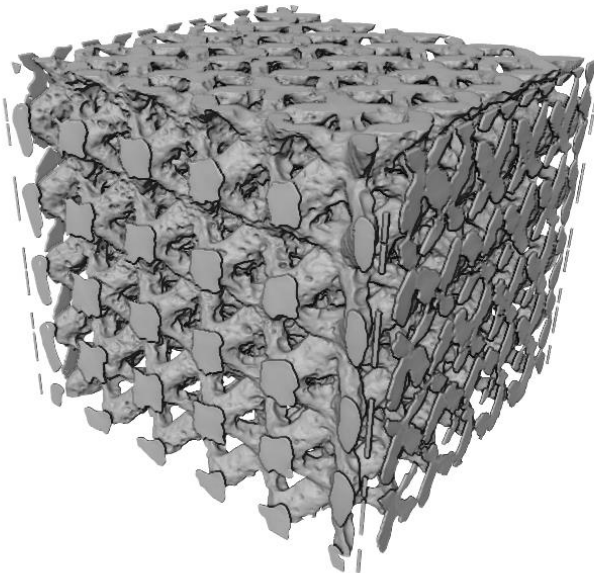


$$\epsilon_{\text{error}} = 0.002 - 0.003\%$$

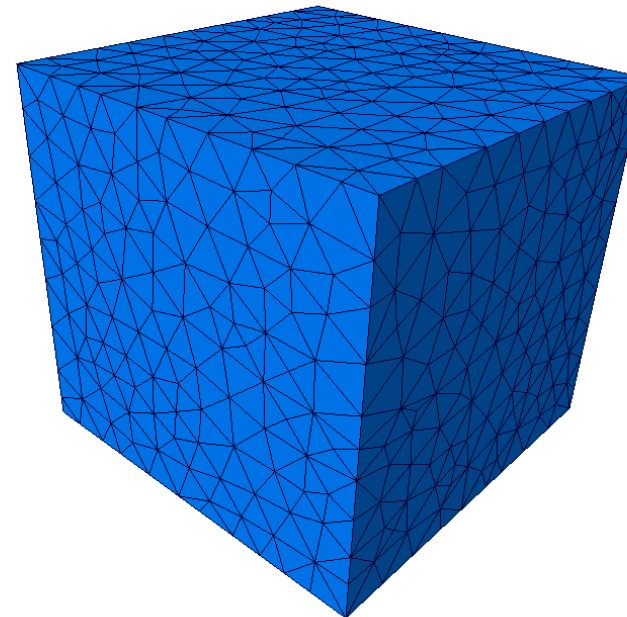
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- **Grid generation:**
  - **Performed automatically by the software.**
  - **It is dependent on the ROI to study.**

BCC Scaffold

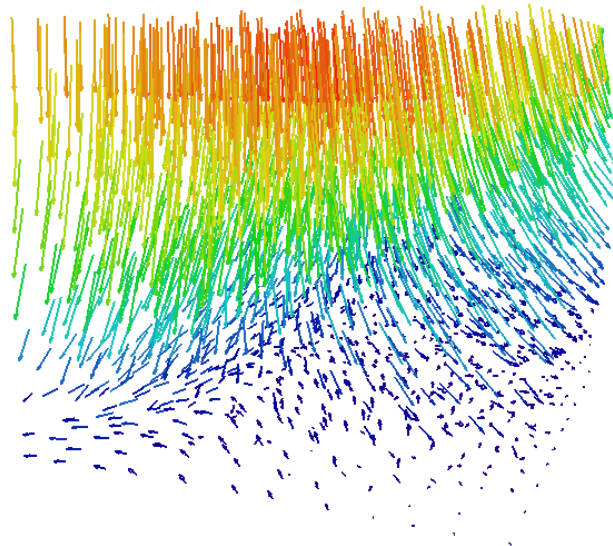


Mesh (170 Cell Size)

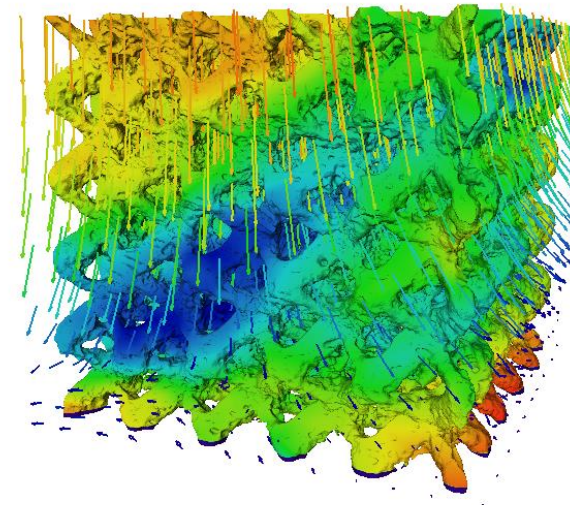


- **Global DVC output:**
  - **Displacement field composed of vectors from each point from their initial positions to their current positions.**
  - **Strain map with a representation of deformations when the material elements were stretched, compressed or sheared.**

Displacements WPB\_0005



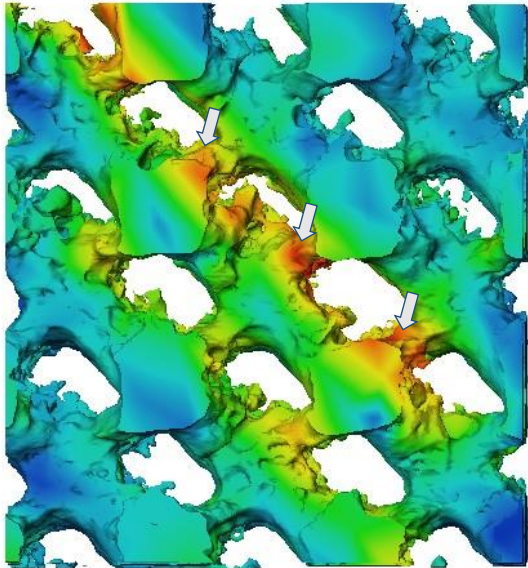
$\epsilon_3$  Strain Map WPB\_0005



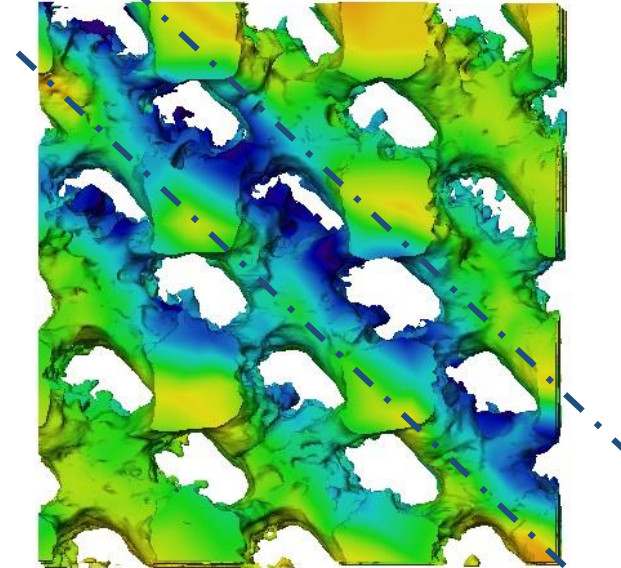
# Degradation of scaffolds

- Maximum,  $\epsilon_1$ , principal stress  $\rightarrow$  Maximum stretching deformation.
- Minimum,  $\epsilon_3$ , principal stress  $\rightarrow$  Maximum compressive deformation.
- The shear band is detected with the maximum compressive deformation.
- Stretching deformation zones acted as catalysts for the shear banding.

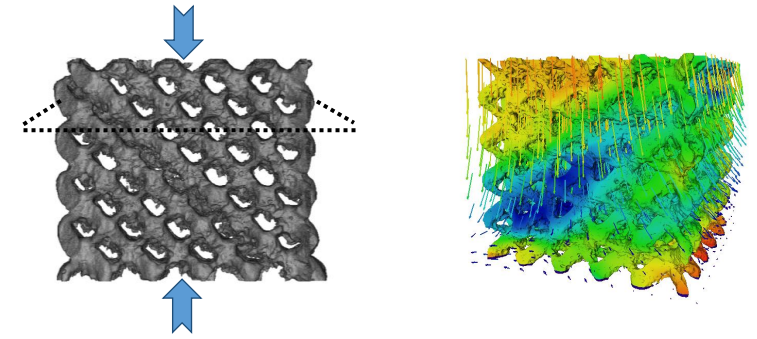
$\epsilon_1$  Strain Map WPB\_0005



$\epsilon_3$  Strain Map WPB\_0005

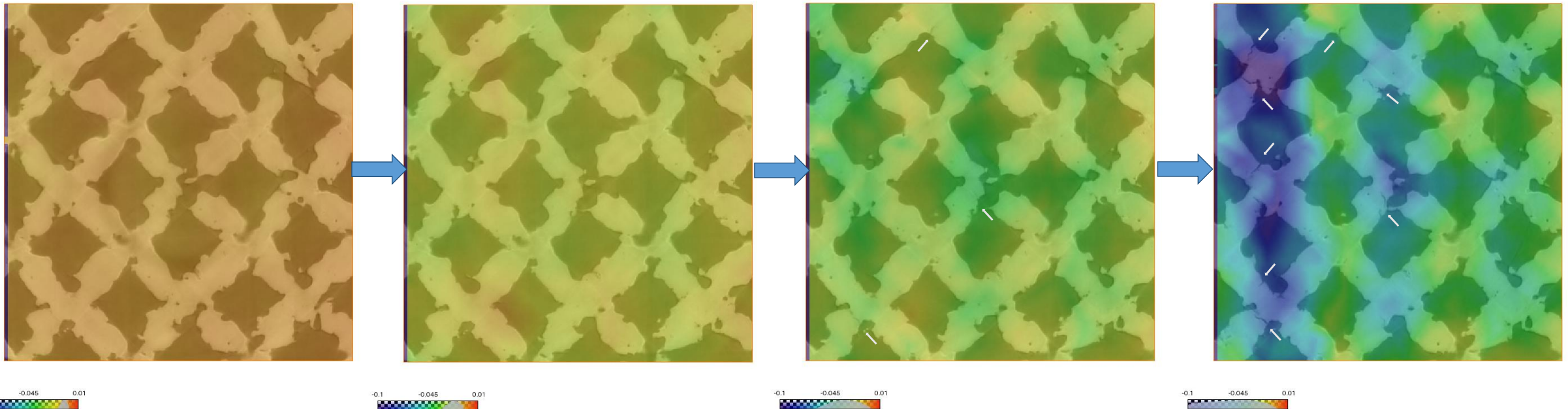


# Degradation of scaffolds



- For a particular slice, the evolution of the strain field with the compression load can be seen.
- Highly deformed regions were located at the intersection of nodes and struts, typical zones for crack initiation.

$\epsilon_3$  Strain Maps



WPB\_002  
( $\epsilon=0.9\%$ )

WPB\_003  
( $\epsilon=2.5\%$ )

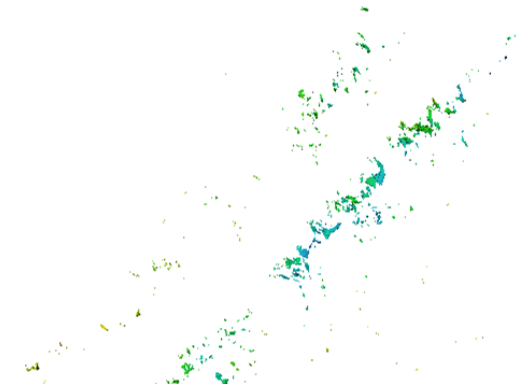
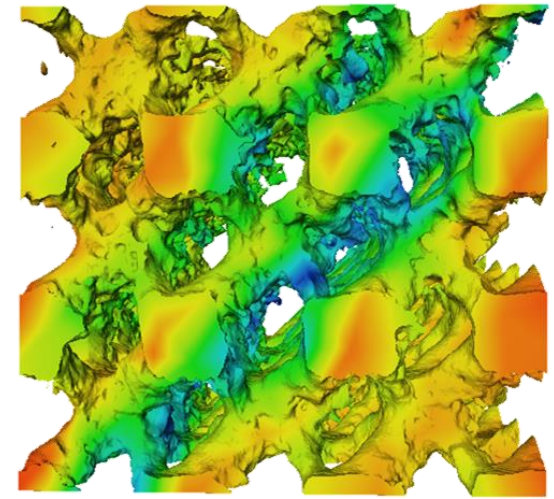
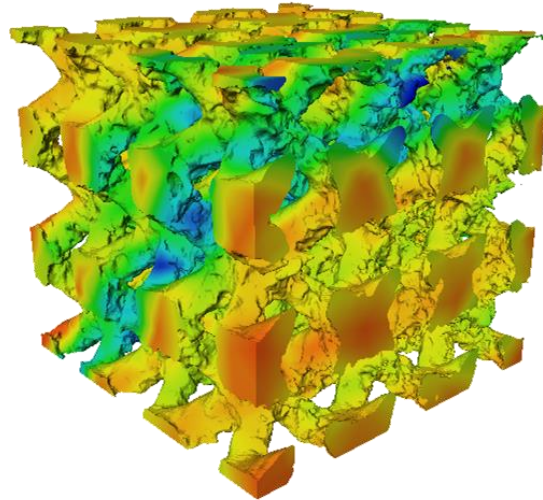
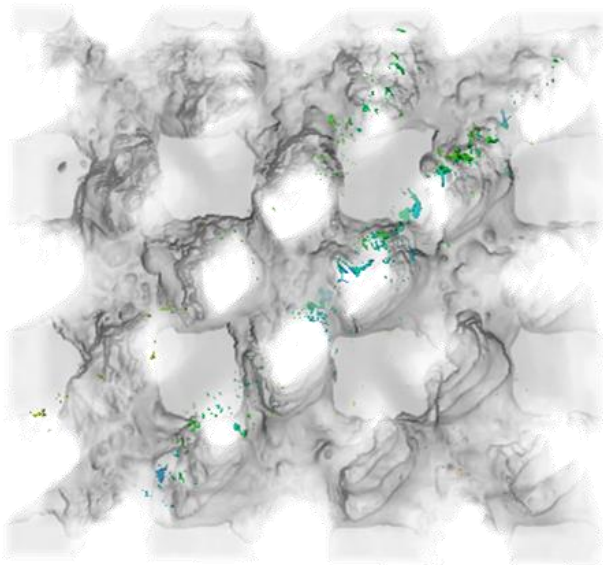
WPB\_004  
( $\epsilon=5.0\%$ )

WPB\_005  
( $\epsilon=6.0\%$ )

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# Degradation of scaffolds

- With the residual from the correlation maps, it was possible to segment and extract the cracks.
- They were located at the diagonal intersections.



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# Summary

# Summary

- **UT is a fast technique, and its precision and detectability can be improved by artificial intelligence.**
- **XCT scans will feed the model.**
- **A massive set of scans is necessary, so the model will identify several types of defects (cracks, bubbles, resin pockets, delamination...).**

# Summary

- **Segmentation process plays a key role in a damage assessment workflow.**
- **Depending on the scan output and reconstruction artefacts several approaches can be applied.**
- **Training the model is the most time-consuming step.**
- **With a well-trained classifier the automated detection will be fast.**

# Summary

- **In-situ scans are a good option to understand the mechanical properties of materials.**
- **The analysis of XCT volumes can be complemented with correlation techniques.**
- **Image and volume correlation allow to identify regions with high deformation, where crack initiation is more probable.**

**Thank you very much!**  
**Questions?**