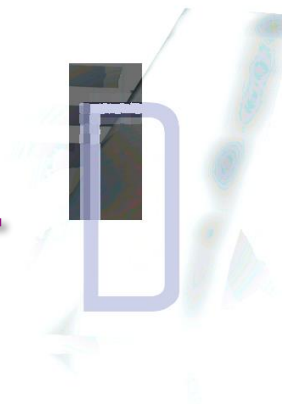




Digital Design strategies using hybrid models to Certify & Manufacture Composite structures

Project presentation

David DUMAS (david.dumas@cenaero.be)



Rationale

Introduction of **new materials or structural concepts** require

- TESTING
- **fundamental understanding at the lower size scale** to feed **methodologies at the macroscopic scale**
- Research entities play a crucial role in the later stages of design and testing
 - Production Method
 - Impact Damage and Tolerance



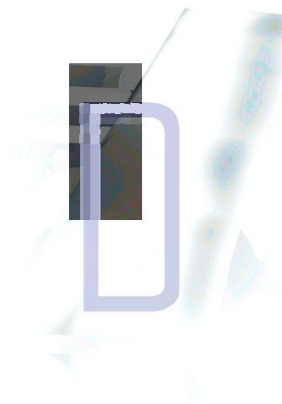
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**Links scales through
machine learning &
hybrid testing**

*[2020 – P. Beaumont,
University of Cambridge]*



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The DIDEAROT Project



*Current active
Advisory Board*



- Horizon Europe project
 - September 2022 – August 2026
 - CINEA agency support
- TRL 2-4 levels
 - Possible outcomes at TRL6 for direct Clean Aviation exploitation

General
presentation

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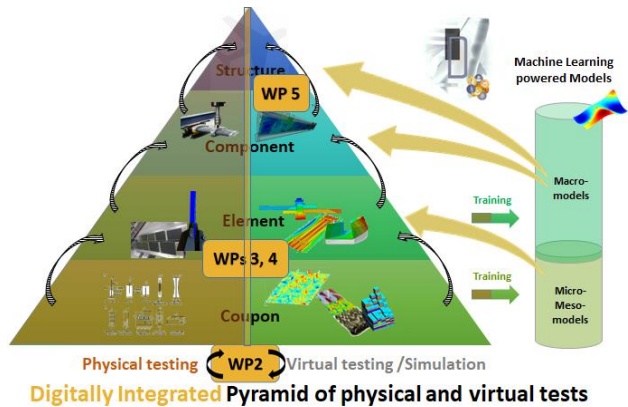
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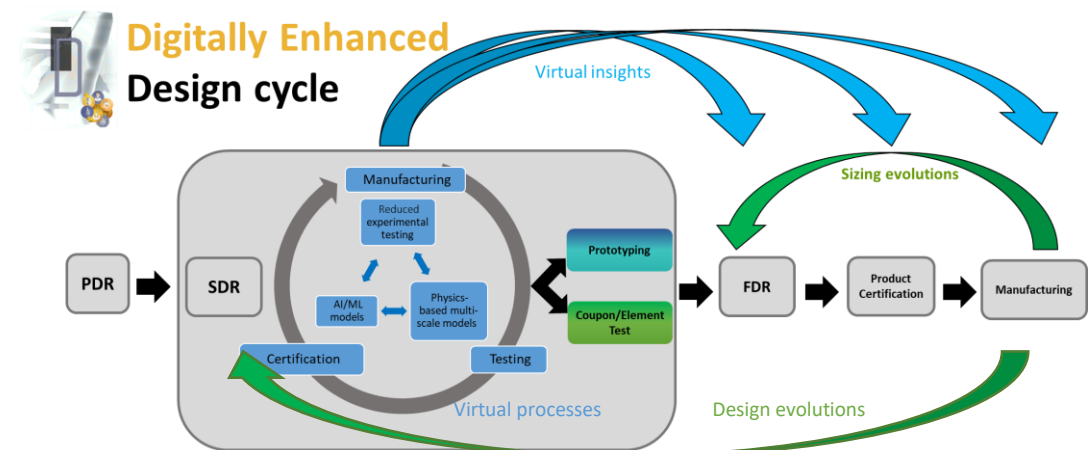
Technical scope & objectives

- Improve the quality & speed of evaluation of complex multi-scale & multi-physics simulation
 - High performance computing environments & optimized codes
 - Adapted models over scales to solve the right problem at the right level



- Build hybrid approaches (machine learning) for response predictions feeding higher levels in the test pyramid

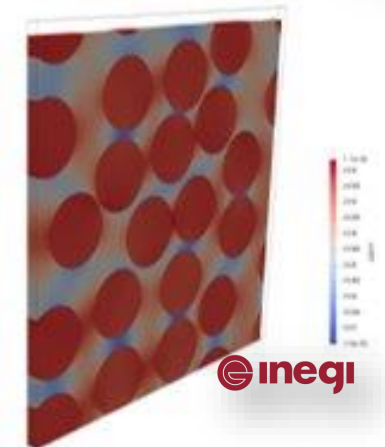
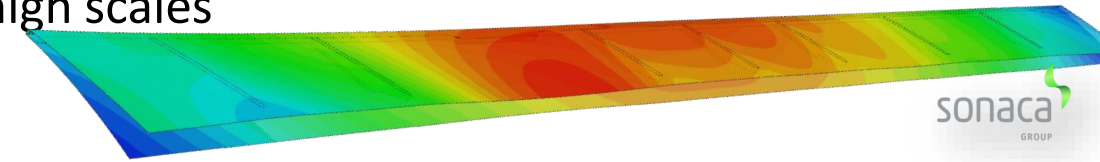
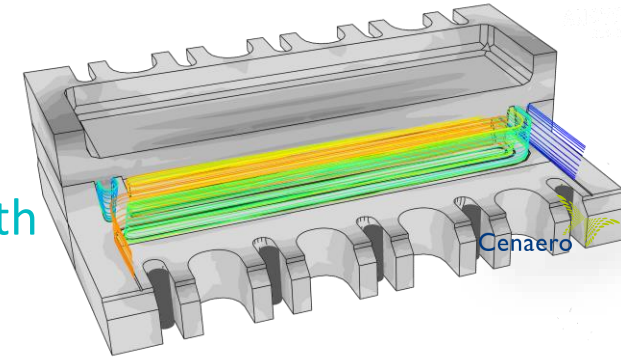
- High end objective
 - Allow simulation to cover manufacturing & vulnerability at the design phase to enable earlier appropriate decisions to limit costly design iterations





Drawing out the blueprint

- **Challenges** of physics through simulation
 - Process simulation
 - Handle coupled fluid-thermal-structural problems through **HPC with enhanced performance**
 - Account for **complex interaction mechanisms** (possibly enrich simulations with experimental results) up to high scales
 - Vulnerability simulation
 - Streamline **model parameter identification** through hybrid methods to bridge the gap between scales
 - Efficiently extend **multi-scale** methods to non-linear response
 - Taking account variability in prediction of response at different scales

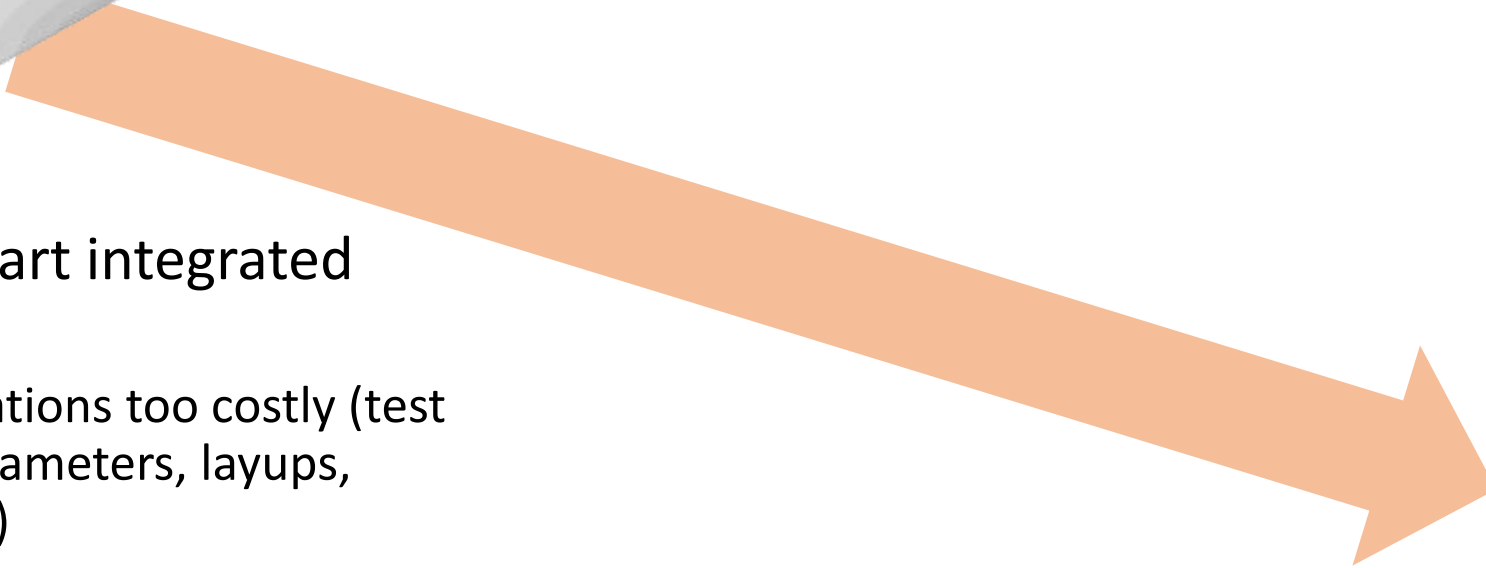




Why perform predictions through machine learning – the case for process simulations

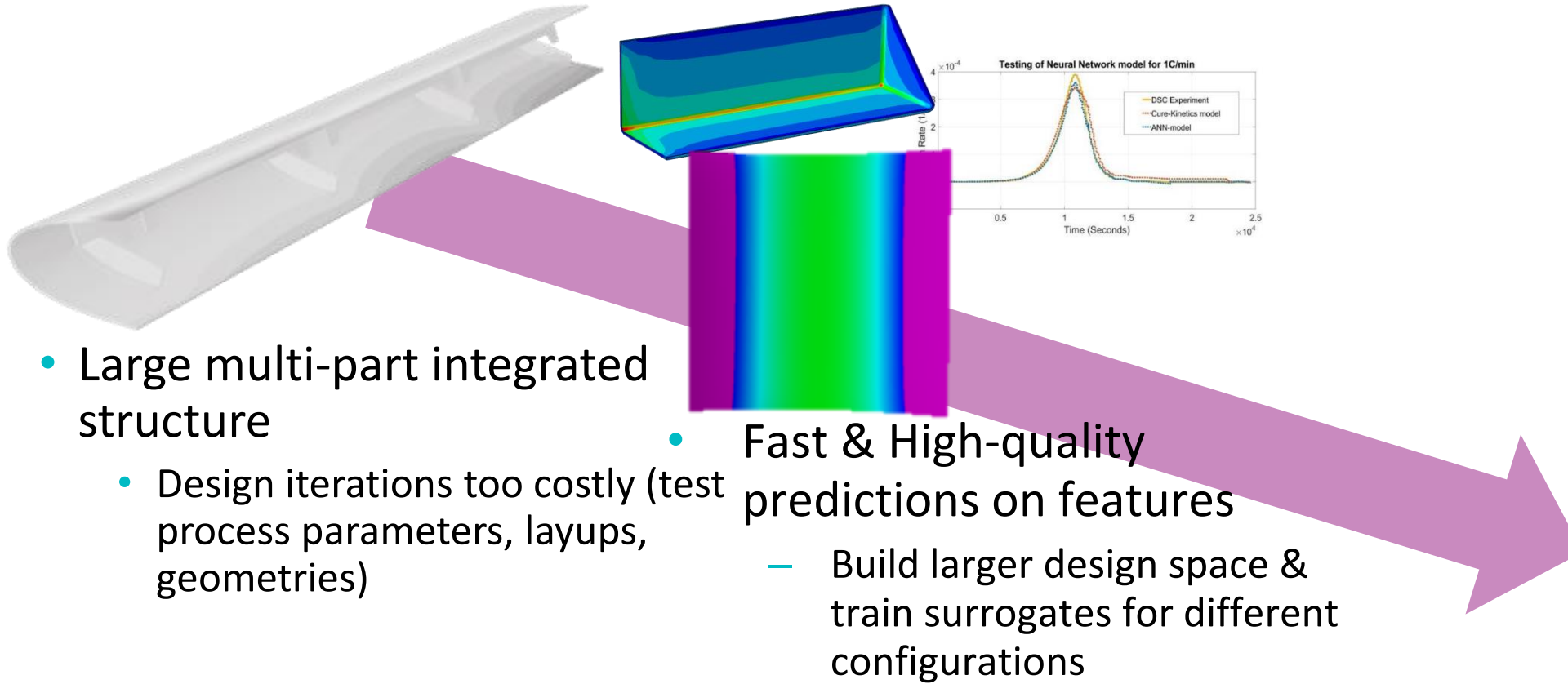


- Large multi-part integrated structure
 - Design iterations too costly (test process parameters, layups, geometries)





Why perform predictions through machine learning – the case for process simulations



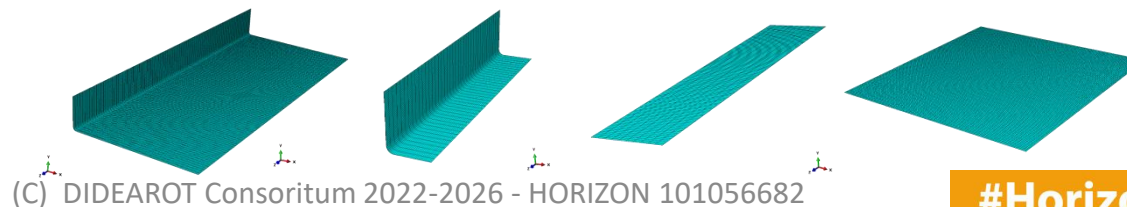
- Large multi-part integrated structure

- Design iterations too costly (test process parameters, layups, geometries)

- Fast & High-quality predictions on features
 - Build larger design space & train surrogates for different configurations



General presentation



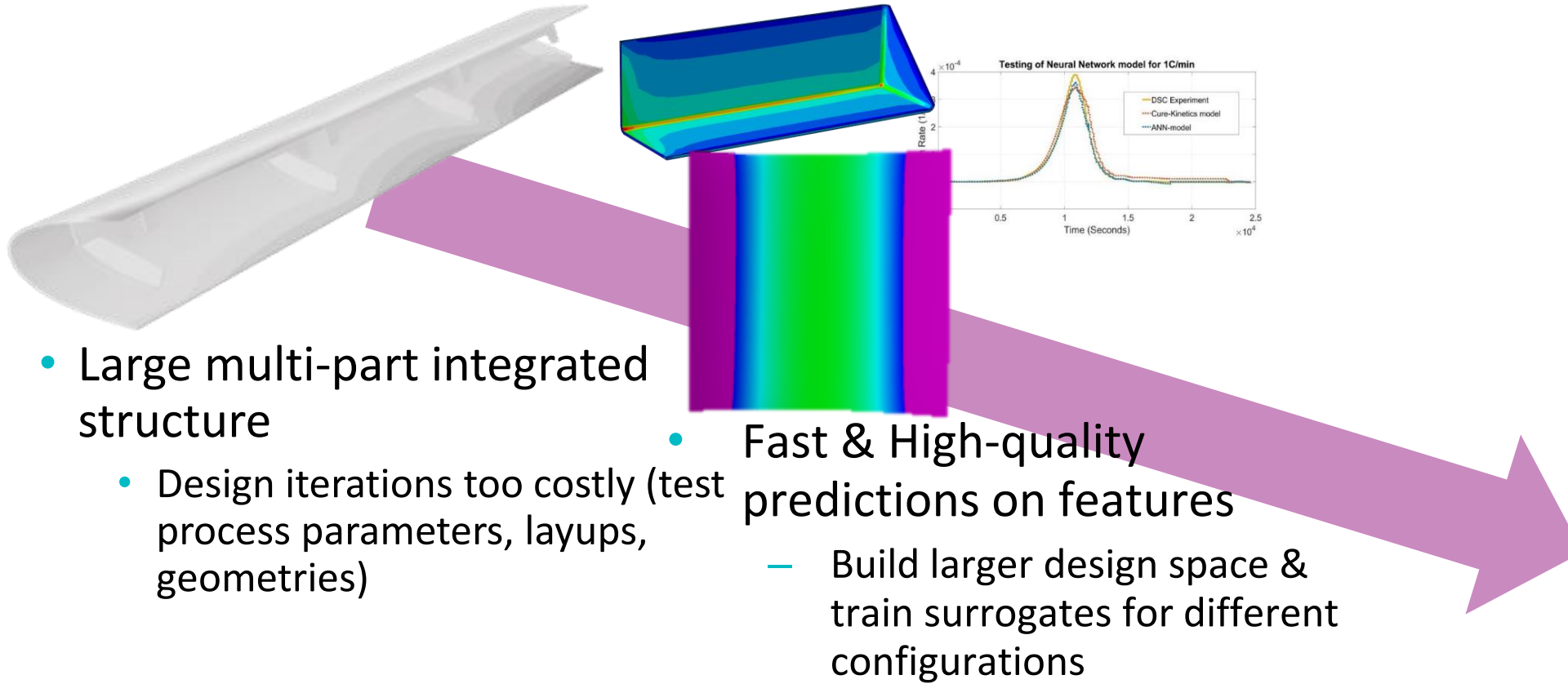
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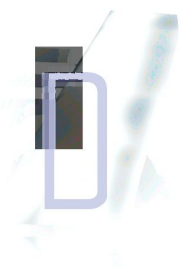
Why perform predictions through machine learning – the case for process simulations



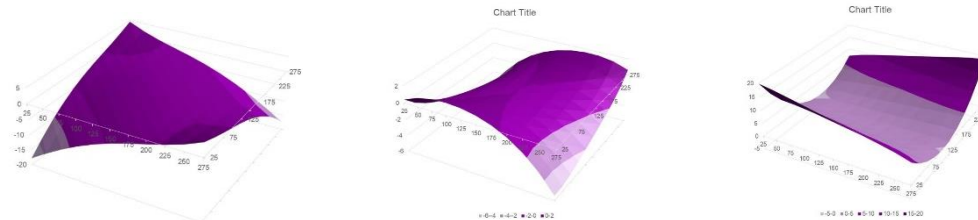
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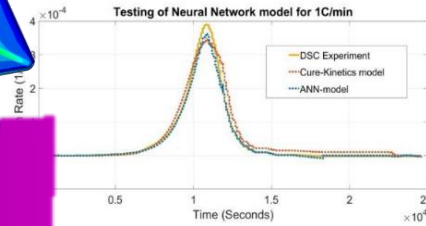
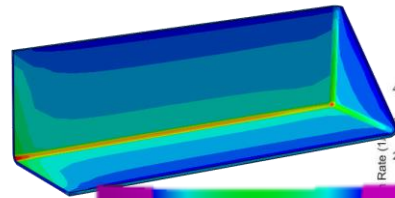
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Why perform predictions through machine learning – the case for process simulations



- Large multi-part integrated structure

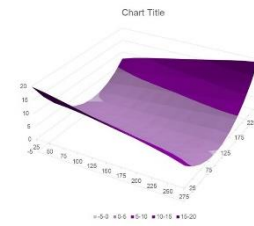
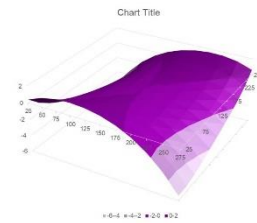
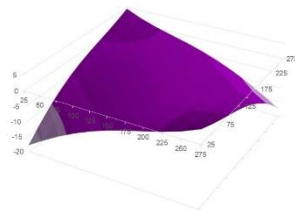
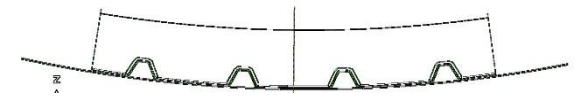
- Design iterations too costly (test process parameters, layups, geometries)

- Fast & High-quality predictions on features

- Build larger design space & train surrogates for different configurations

- Build large scale predictions

- Break down large parts into features
- Test training of assembly of surrogates from large scale simulations
- Find processing trends, UQ & build robust configurations



Why perform predictions through machine learning – the case for process simulations



- Large multi-part integrated structure

- Design iterations too costly (test process parameters, layups, geometries)

- Fast & High-quality predictions on features

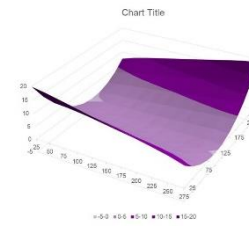
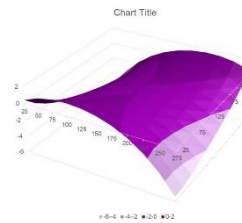
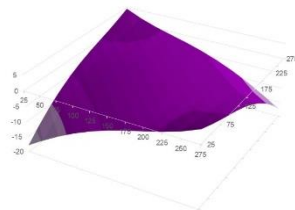
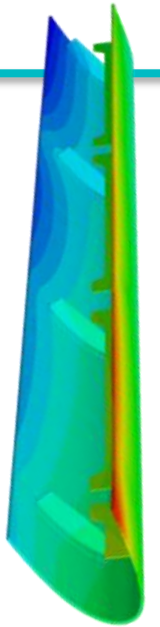
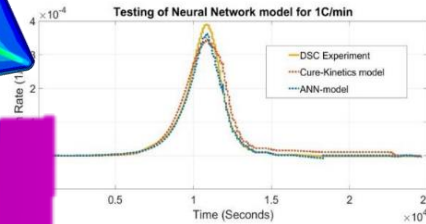
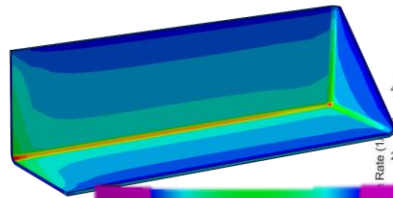
- Build larger design space & train surrogates for different configurations

- High-fidelity large-scale predictions

- Validate processing route on 1-2 configurations through high fidelity simulation

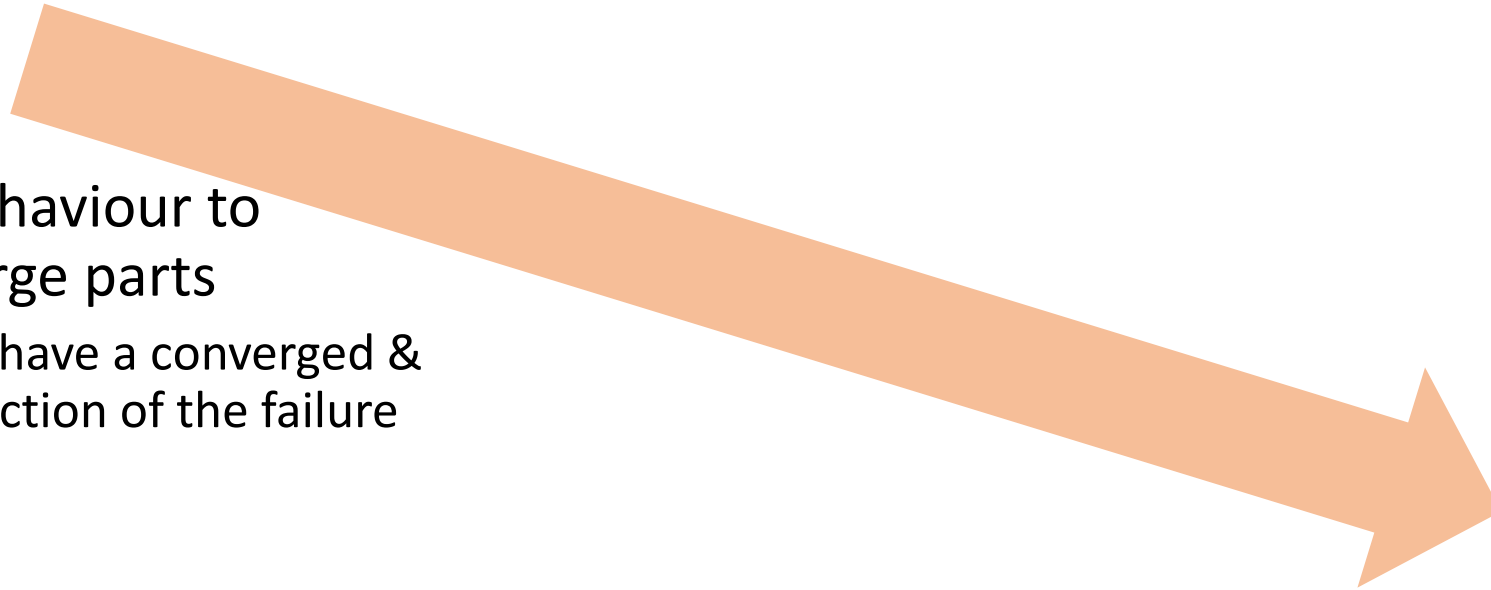
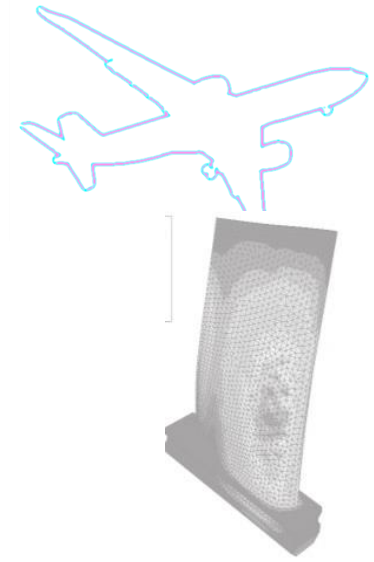
- Build large scale predictions

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- Test training of assembly of surrogates from large scale simulations
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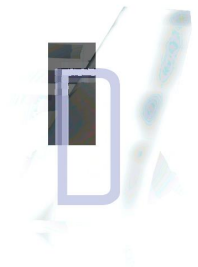


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Why perform predictions through machine learning – the case for process simulations



- Composite behaviour to damage on large parts
 - Difficulty to have a converged & robust prediction of the failure analysis



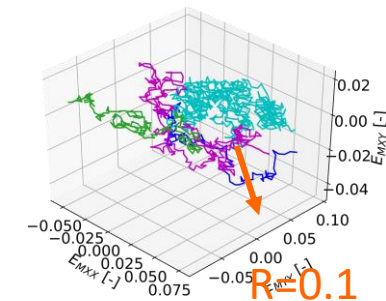
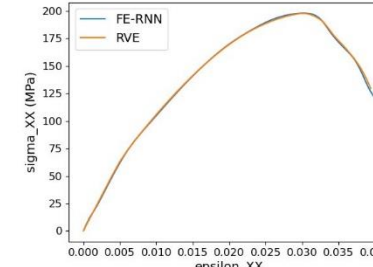
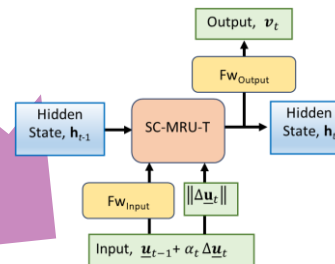
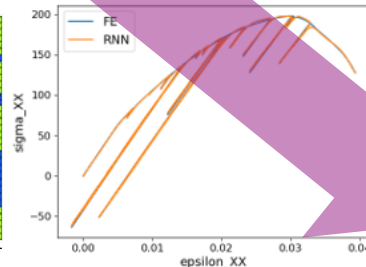
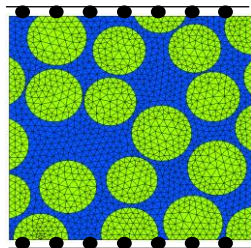
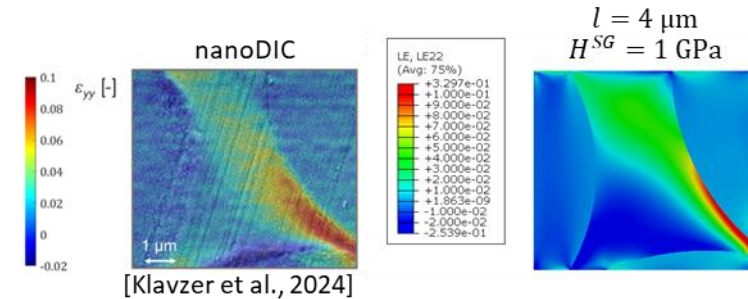
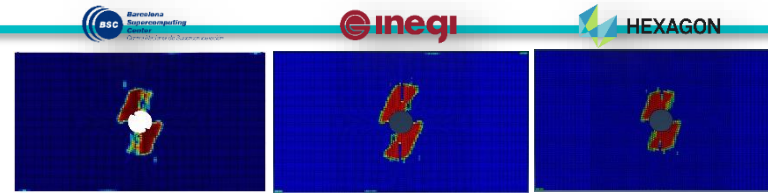
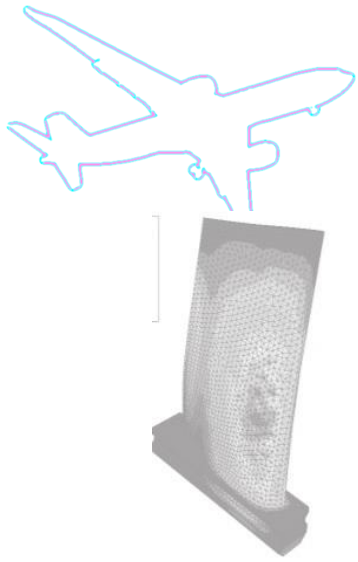
Why perform predictions through machine learning – the case for process simulations

- Composite behaviour to damage on large parts

- Difficulty to have a converged & robust prediction of the failure analysis

- 3D elasto-plastic RNN with damage to predict evolution on random load paths

- Damage models applicable to many load cases



General presentation

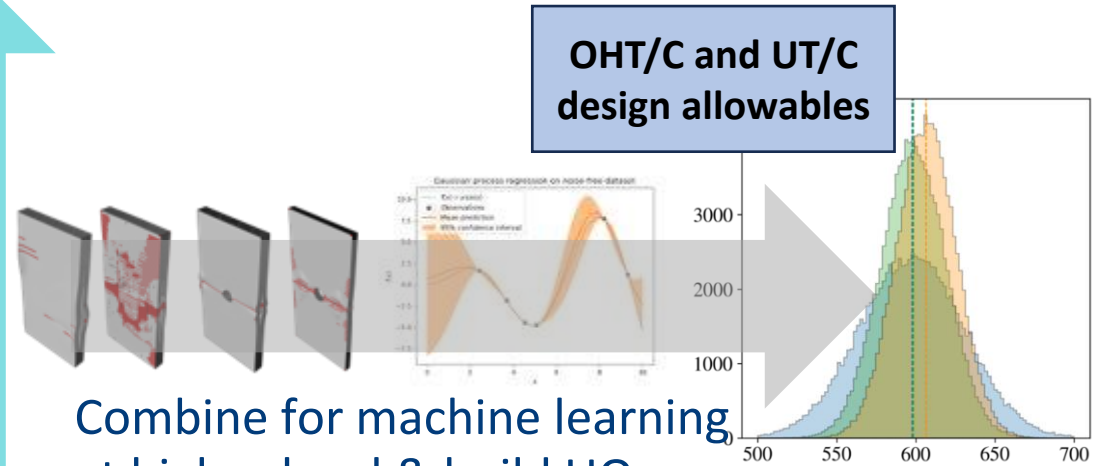
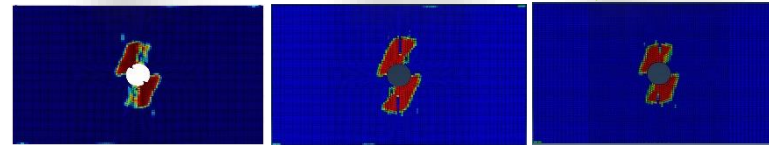
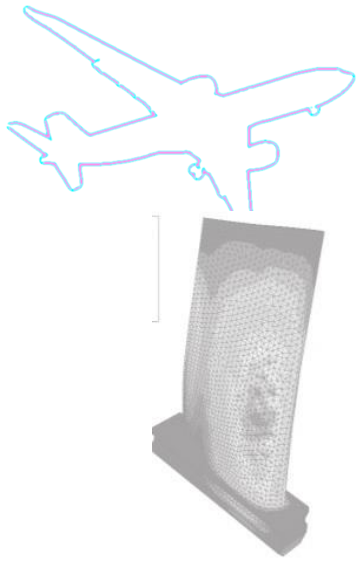
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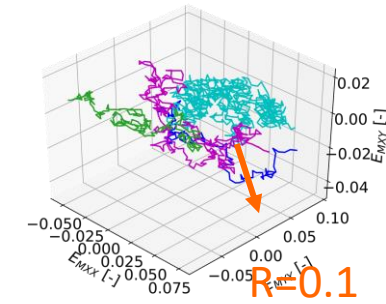
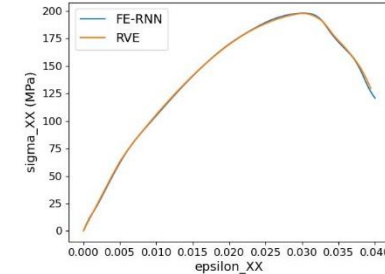
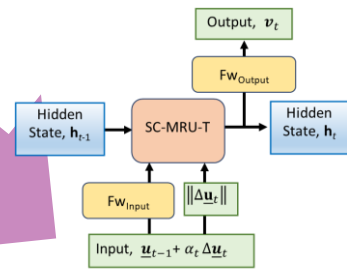
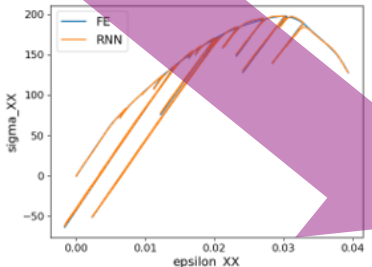
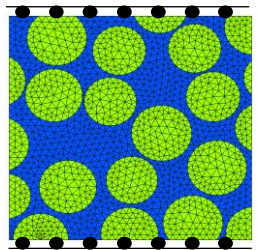
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Why perform predictions through machine learning – the case for process simulations

- Composite behaviour to damage on large parts
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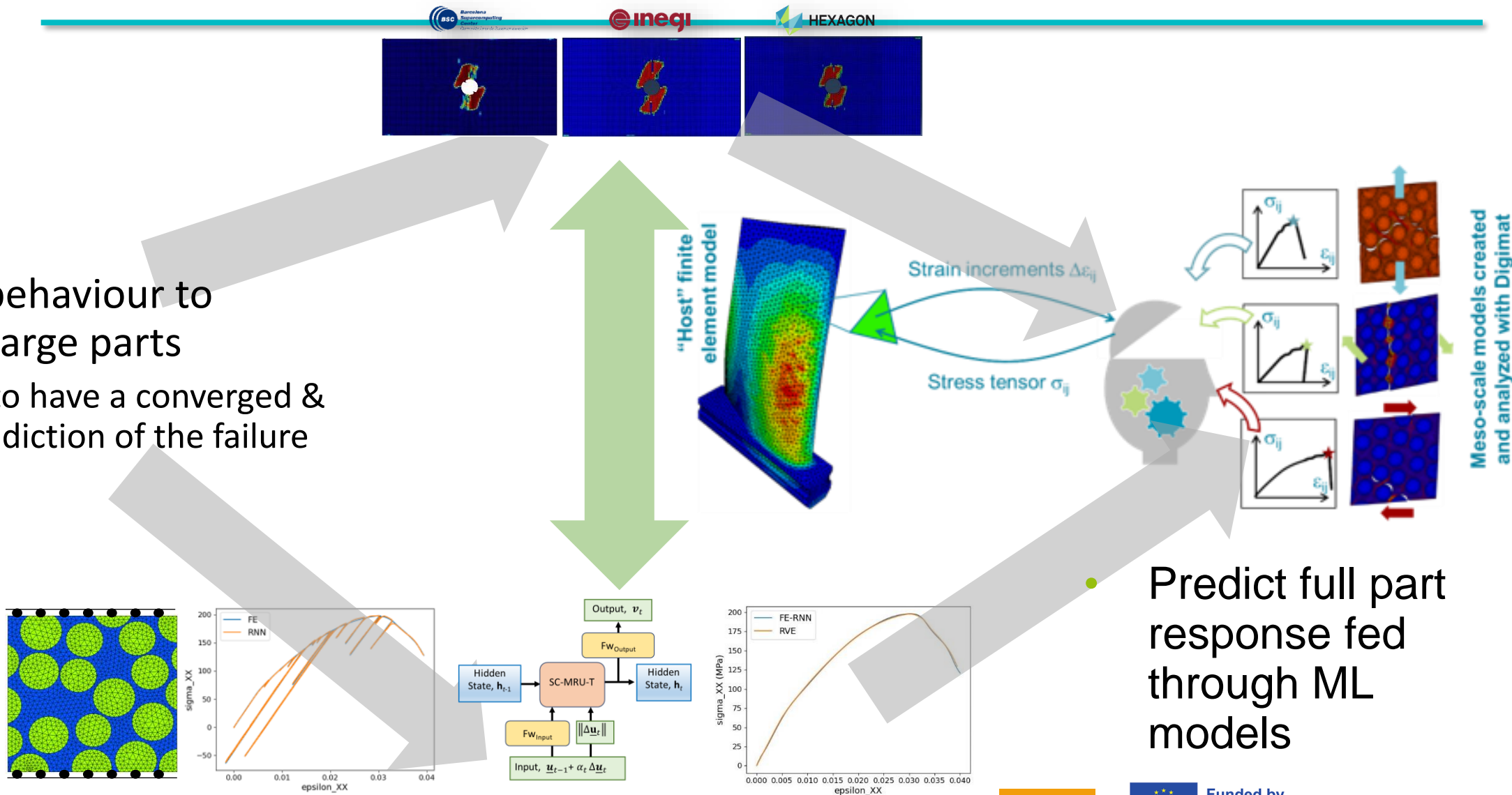
Combine for machine learning at higher level & build UQ response



General presentation

Why perform predictions through machine learning – the case for process simulations

- Composite behaviour to damage on large parts
 - Difficulty to have a converged & robust prediction of the failure analysis

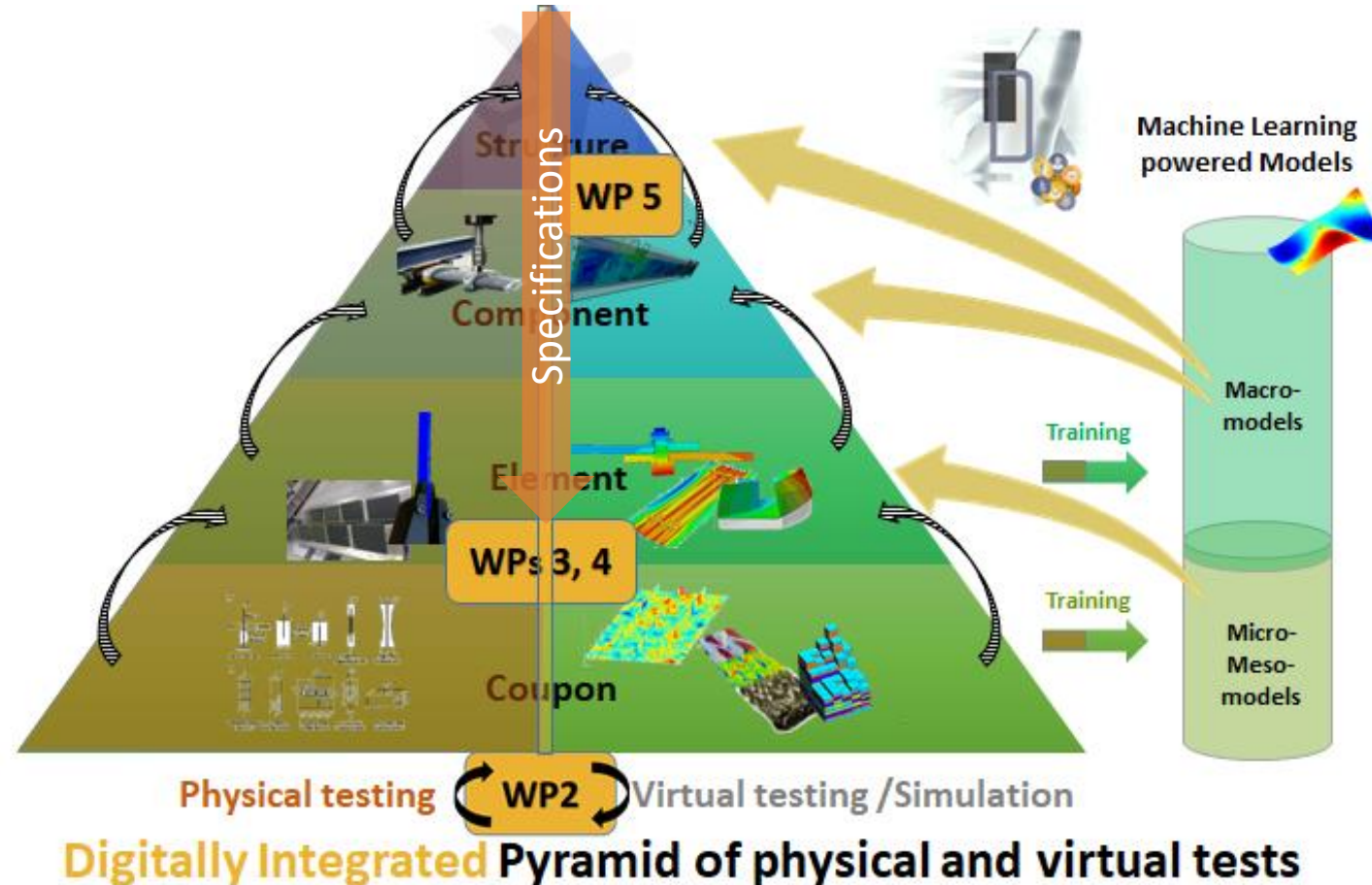


• Predict full part response fed through ML models



Simulation enhanced validation over scales

- Approach in building a hybrid testing pyramid





Path to exploitation for integration & certification

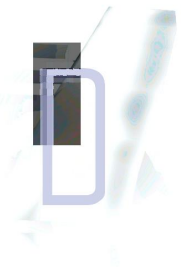
- Perform assesment of the numerical approaches
 - Performance benchmarks & Performance Optimisation of codes
 - Verification of the portability of the numerical approaches in other architectures
- Optimization of algorithm & computational efficiency of codes for HPC
- Account for uncertainty for the implementation of robust simulation





Upcoming activities

- DIDEAROT project will be carried out until end of 2026
- Regular bi-monthly internal technical progress meetings for partners
- Advisory board meetings at annual intervals
 - M36 (2025)
 - M48 (2026)
- Workshop initiatives near end of project
 - One with more focus on Academics & HPC
 - One focused on applications & transfer industry
 - Link to network of projects in the same call & Clean aviation
- Feedback and interactions with Advisory board based on specific requests
- Build exploitation plan with industrial partners (demonstrators & software)



Thank you for your attention



[linkedin.com/company/didearot-horizon-Europe](https://www.linkedin.com/company/didearot-horizon-Europe)



www.didearot-project.eu



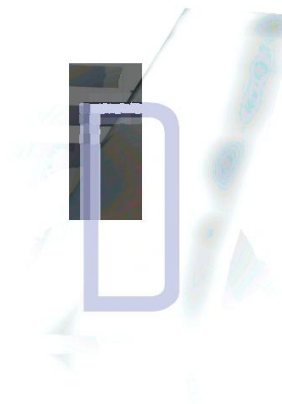
<https://gitlab.uliege.be/didearot>



This project has received funding from the European Union's Horizon Europe research and innovation programme under grant agreement No 101056682.

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The present research benefited from computational resources made available on Lucia, the Tier-1 supercomputer of the Walloon Region, infrastructure funded by the Walloon Region under the grant agreement n°1910247.



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