

A wireframe model of an airplane, rendered in a glowing cyan color, set against a dark blue background with bokeh light effects. The model shows the fuselage, wings, and tail section.

More efficient fatigue evaluation of composite structures by leveraging machine learning and surrogate modelling

John-Alan Pascoe (TUD)

EMMC – MatCHMaker Workshop October 24th



Funded by
the European Union

Summary

p.03

Project background

p.05

Project overview

p.09

Project consortium

p.12

Work Breakdown Structure,
interrelations and timeline

p.14

Techniques developed

p.19

EMMC & EMCC

p.22

Cross-fertilisation

p.24

Contact Points

01

Project background

Motivation



Project background

Motivation

Situation

- › Composite materials crucial for sustainable future
 - Aircraft 50% or more out of FRP
 - Wind energy sector with larger turbines
- › Extreme loads and stress cycles during product lifetime
- › Damage tolerance needed to extend the lifespan

Challenge

- › Lack of fundamental knowledge regarding damage growth process under cyclic loading
- › Complex interaction of fibres, fibre orientation, and matrix materials
- › Every new design requires multiple tests at various scales
- › Imperfect solutions: high-safety factors, 'no growth' criterion

Consequence

Composites products are

- › Expensive
- › Laborious
- › Inefficient to manufacture

Need

A better understanding of the fatigue characterization at a meso-scale, so that we can scale up the gained insights to larger structures

Project background

Problem statement

For composites: each lay-up is a new 'material'

Practical consequence: limitation of design freedom to qualified lay-ups

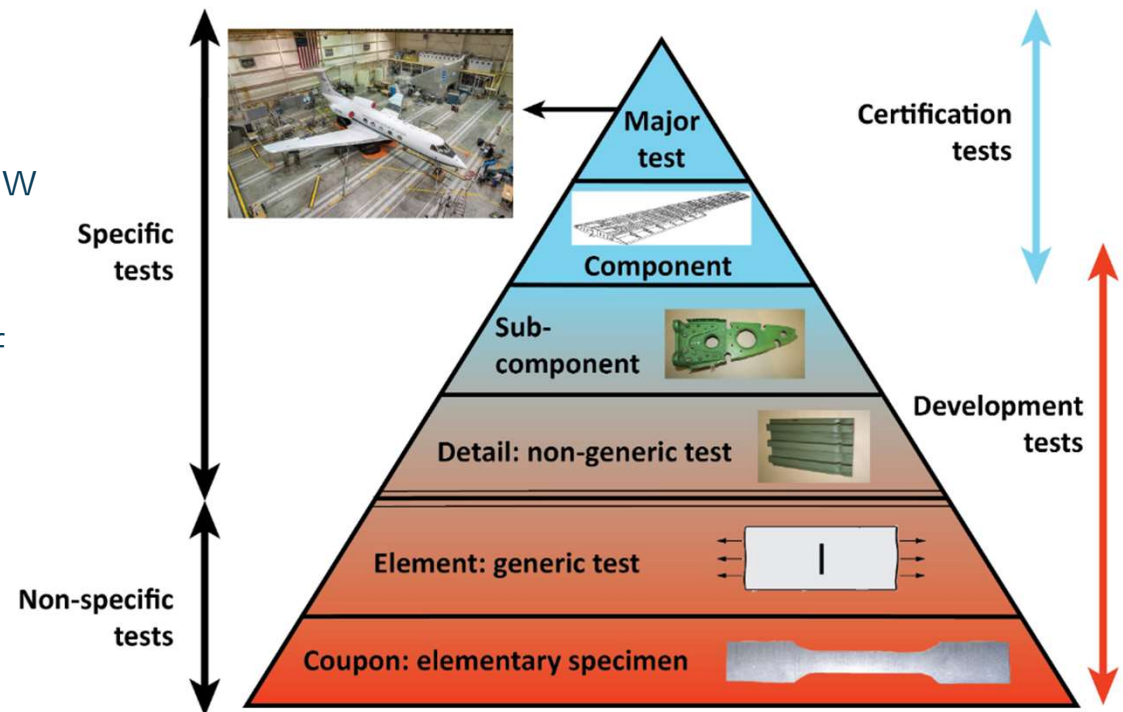


Image: Alderliesten (2018), Introduction to Aerospace Structures and Materials, TU Delft Open, CC-BY-NC-SA 4.0, <https://doi.org/10.5074/t.2018.003>

02

Project overview

Objective, approach, and impact



D-STANDART objective

Project intention

What?

To develop **rapid** methods to **characterise fatigue damage** in composites and **sustainability of composite** supply chains; and thereby **model the durability** and sustainability of large-scale composite structures with arbitrary layups **under realistic conditions** (operating loads, application environment, and impact of manufacturing defects)

How?

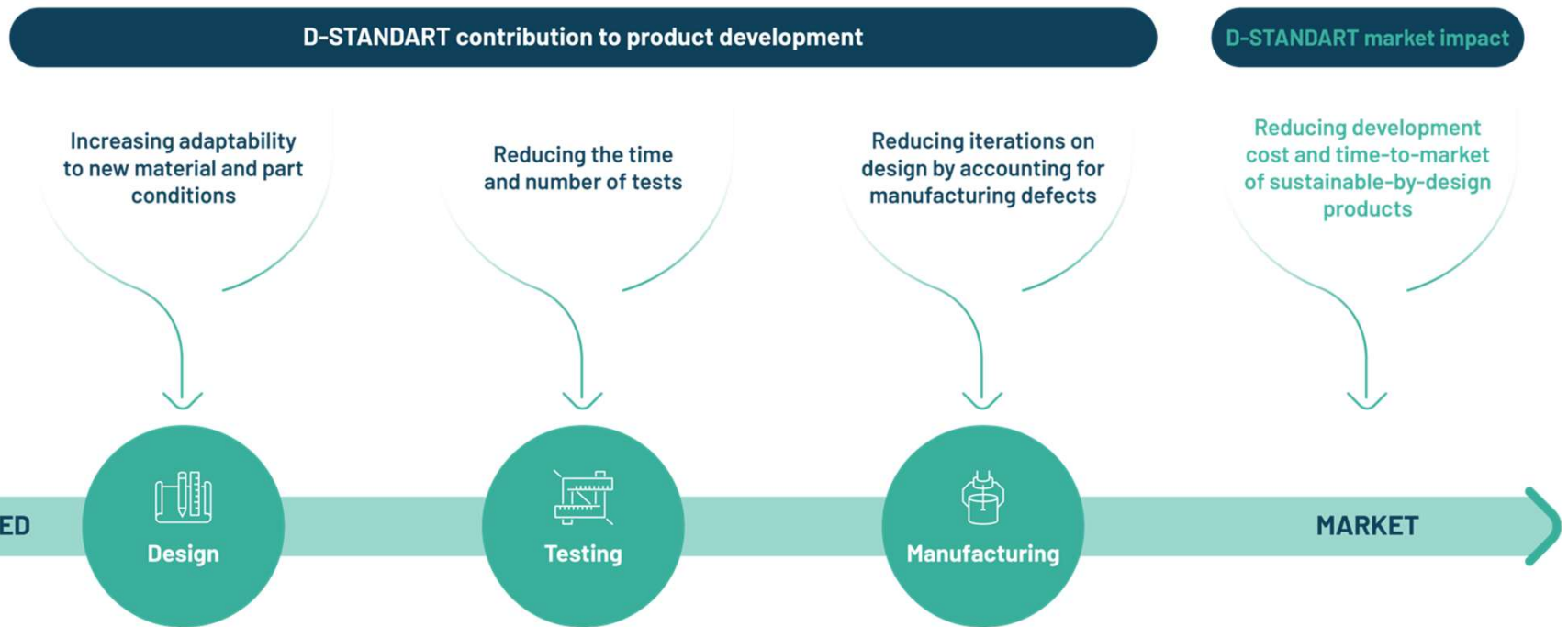
- › **Through minimal and accelerated testing** of generic specimens
- › Transferring the results of the experiments to large-scale structures using **artificial intelligence and machine learning**

Effect?

Enabling **reduced time-to-market, material waste, and increased lifespan** of composite products in the **aerospace and wind energy** industries

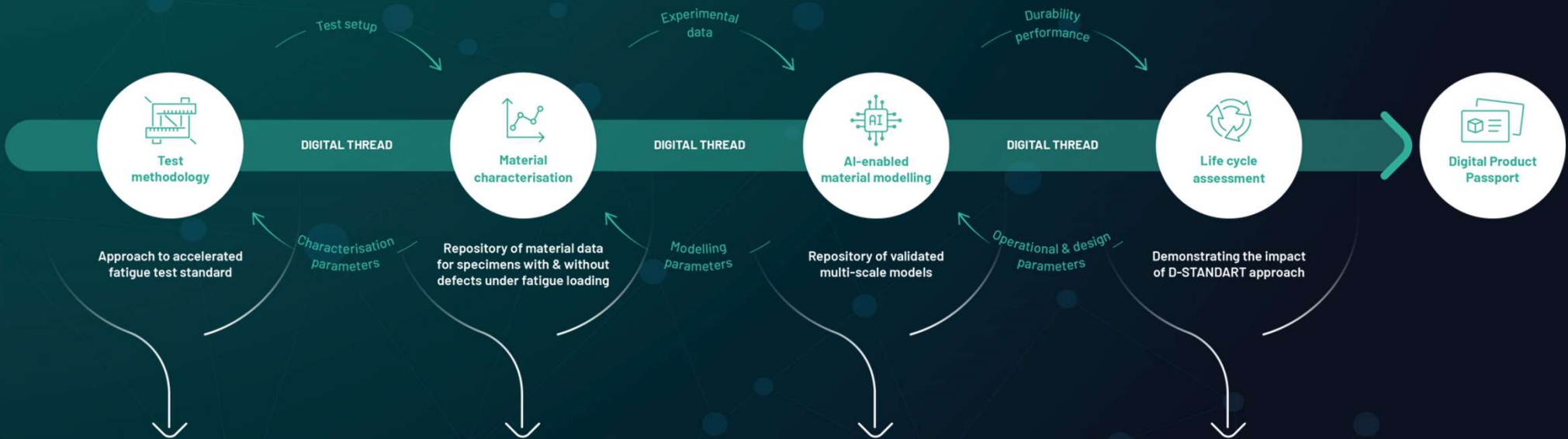
D-STANDART contribution and impact

Addressing the “what” and “effect” of the objective



D-STANDART approach

Addressing the “how” of the objective



IMPACT

Reduced design & testing times, more reliable fatigue modelling, higher fatigue performance, reduced time-to-market, sustainable-by-design products

03

Project consortium

Participants and beneficiaries



Chapter 3

D-STANDART in a nutshell

Consortium, Effort in PM, Duration



NLR

71.4 PM

UNIVERSITY OF TWENTE.

UT

79 PM



TUD

97.5 PM



HMI-D
and affiliates

41 PM



LUP

25 PM



SUZ

42 PM



UB

91.2 PM



NCC

70,5 PM



ICO

30.5 PM

Total:

548.1 PM

GA: 101091409

Start: 1 January 2023

Duration: 36 months

6 Beneficiaries, **4** Affiliates

3 Associated Partners

7 European countries

5 RTOs

1 IND

2 SMEs

Advisory Board Members

Initial list



Rolls Royce



Fokker
Aerostructures



Siemens Gamesa



Embraer



Coexpair (SME)

04

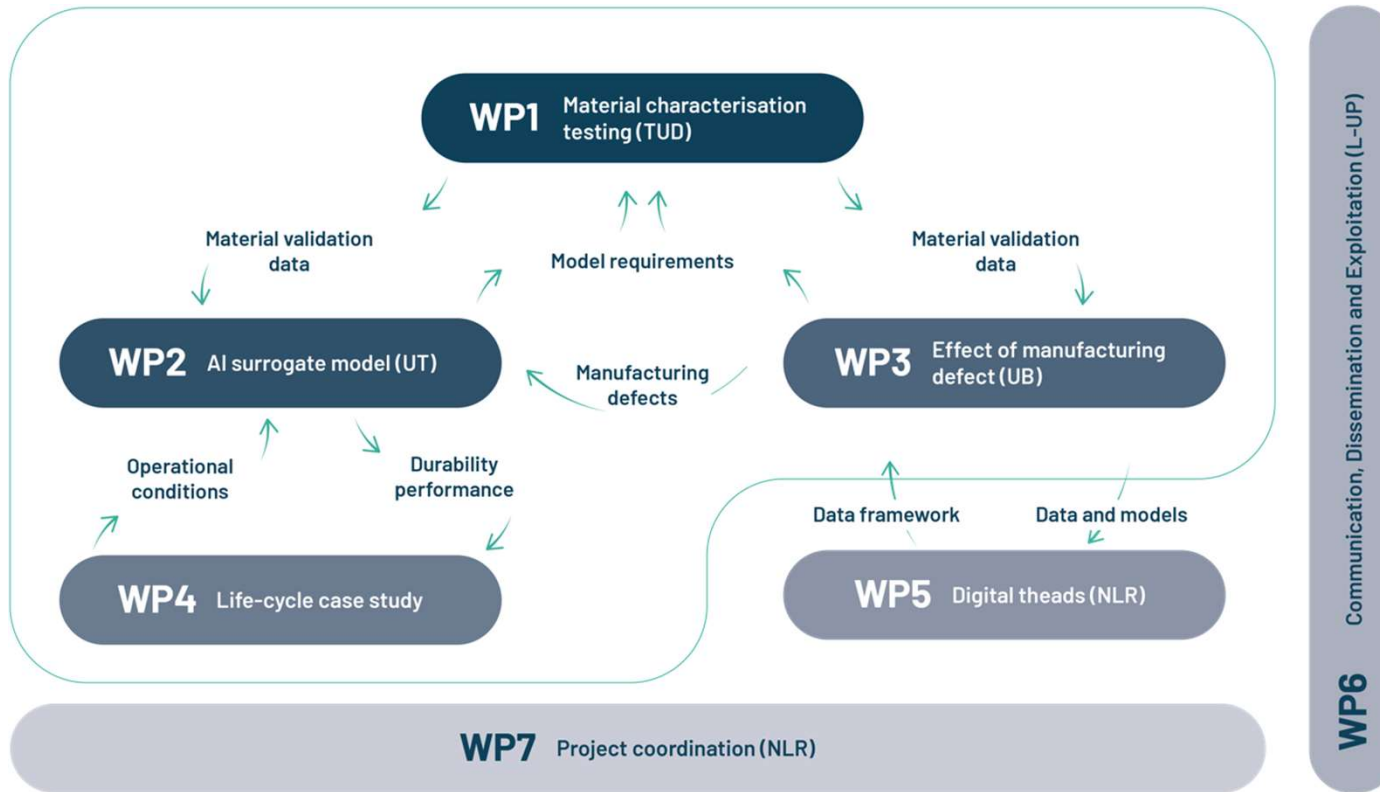
Work Breakdown Structure, interrelations and timeline

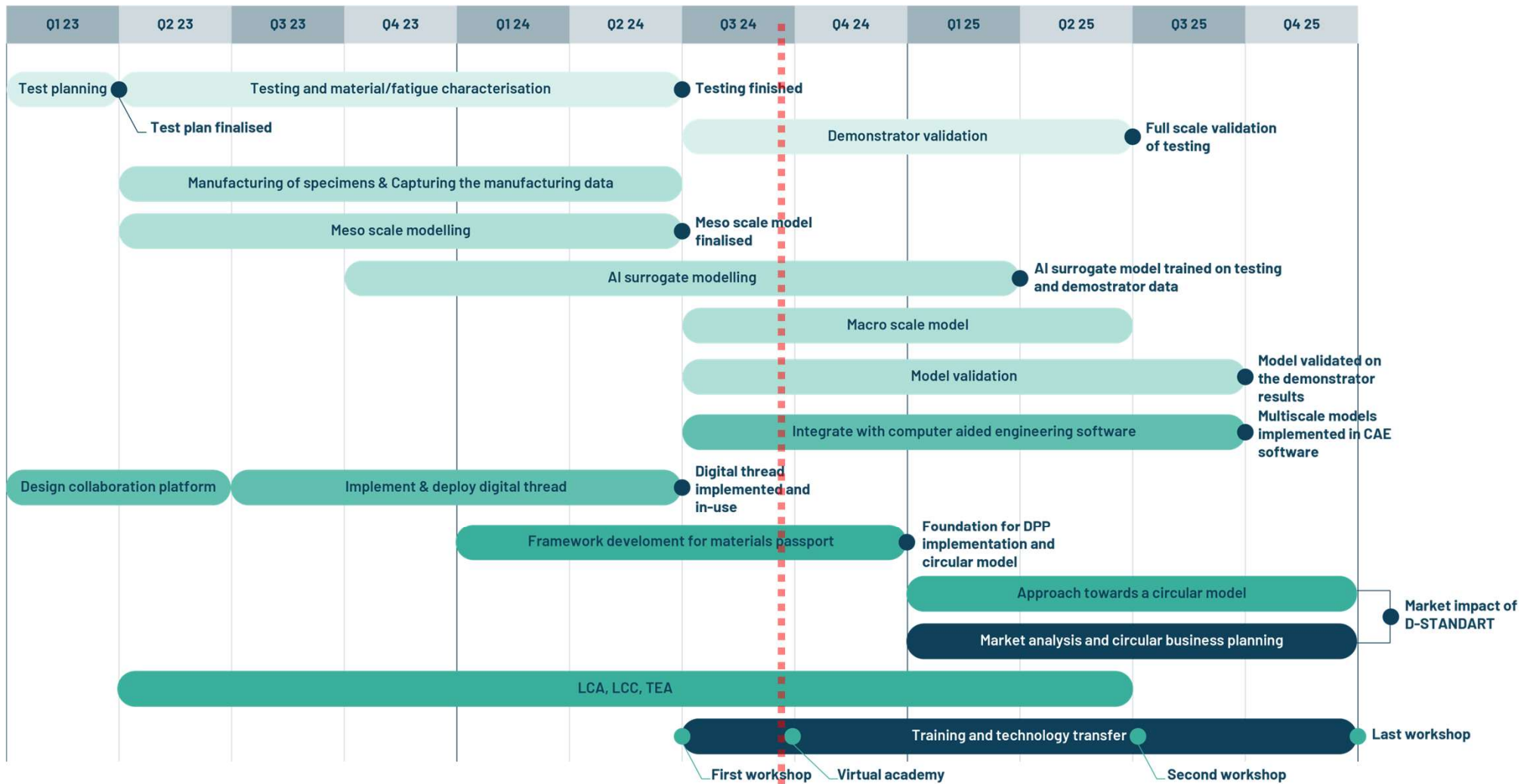
Breakdown of activities



D-STANDART WBS

And interrelations





05

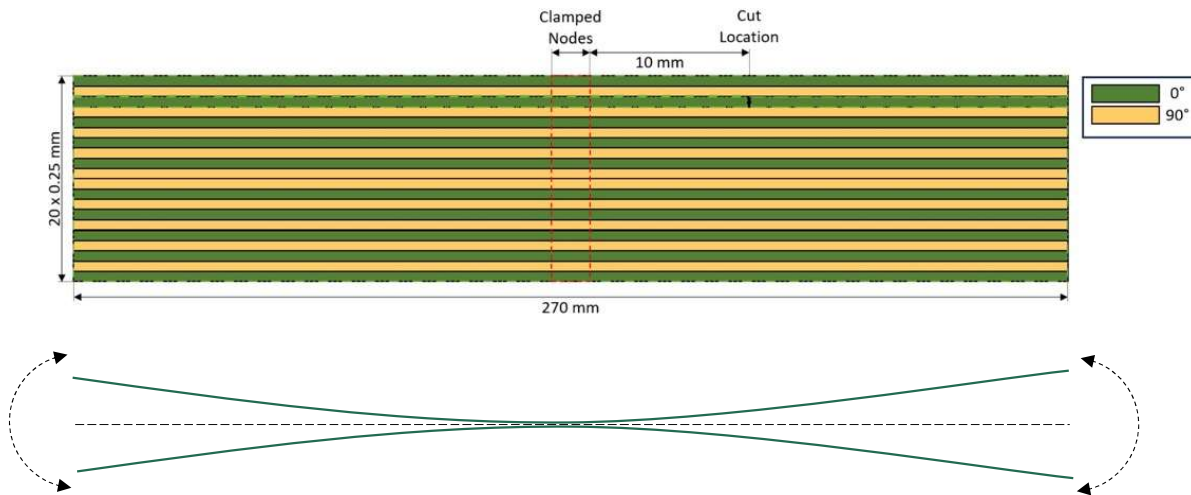
Method Development

Approach & first results



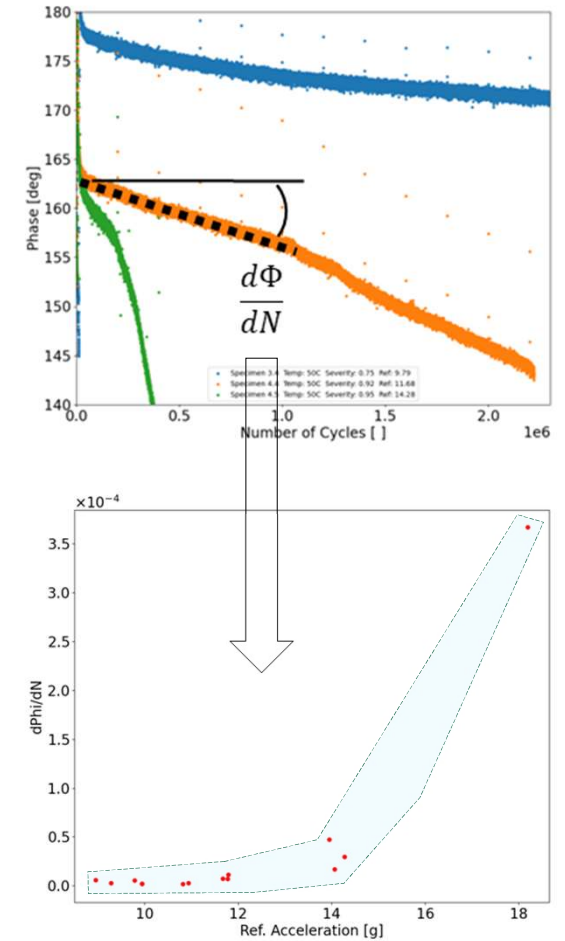
Accelerated testing

High frequency testing



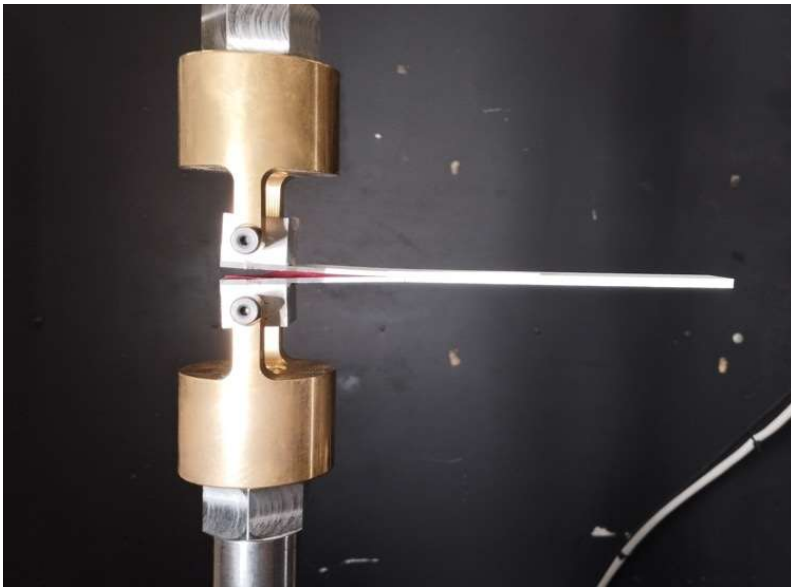
Strong correlation response phase drop, damage propagation, and temperature increase.

Challenge: correlate to traditional fatigue metrics & design tools

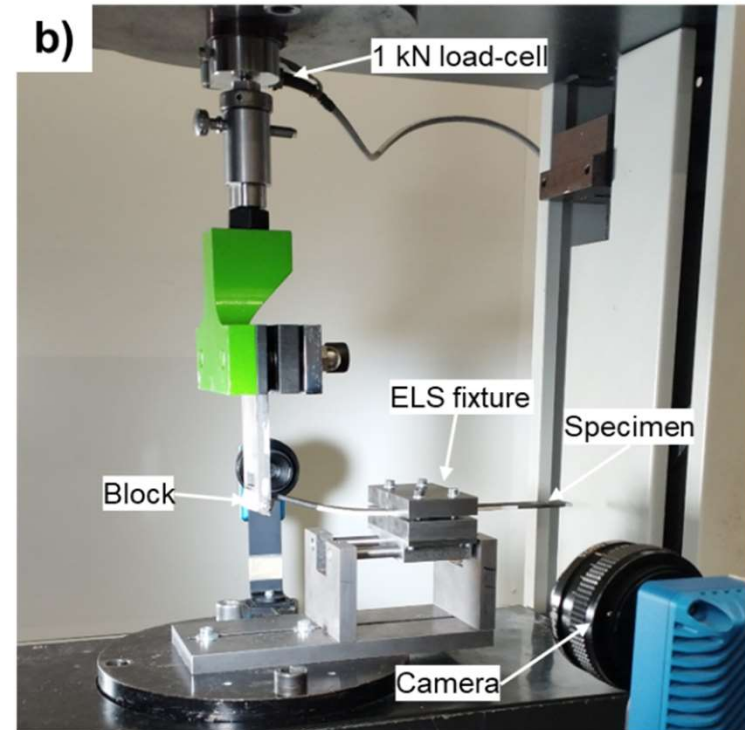


Fibre orientation effect

Based on existing ASTM / ISO standards



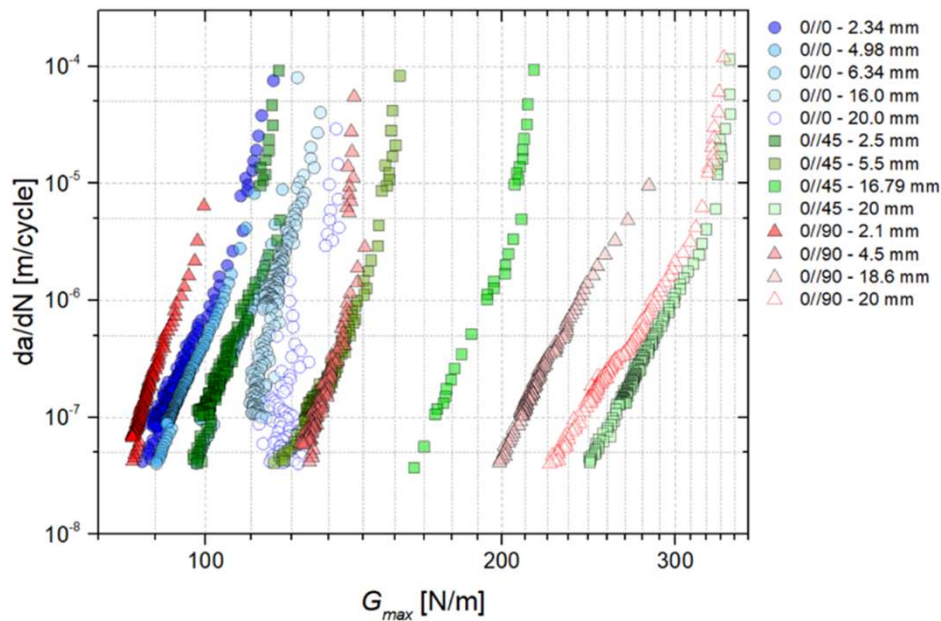
Mode I



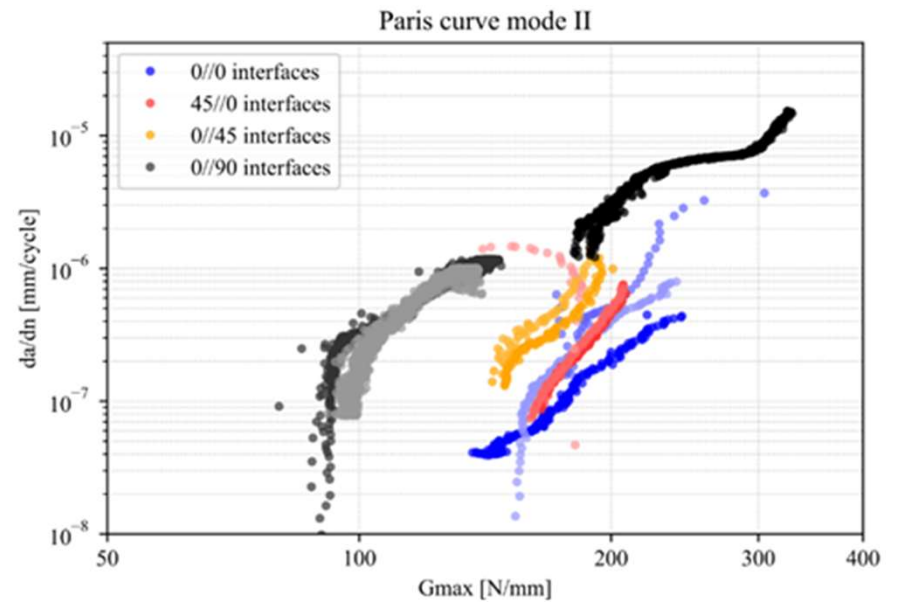
Mode II

Fibre orientation effect

Based on existing ASTM / ISO standards



Mode I: Mainly effect at longer pre-cracks -> more fibre bridging



Mode II: Strong effect also for short pre-cracks

ANN to capture fibre orientation effect

Ongoing trials

ECCM21
02-05 July 2024
Nantes - France

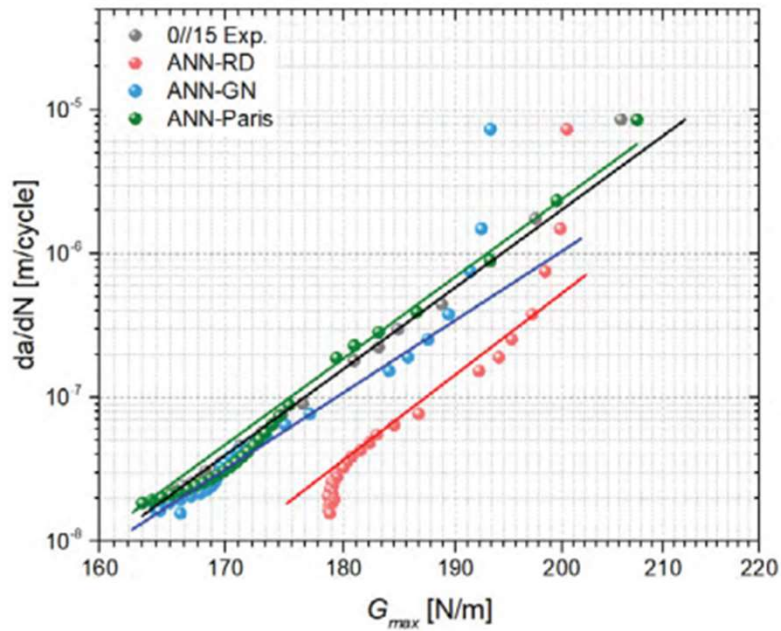
Proceedings of the 21st European Conference on Composite Materials
Volume 8 - Special Sessions

571
1420

Identifying the Most Effective Data Processing for Fatigue Delamination Growth in FRPs: Insights on Artificial Data Simulation

Francisco Maciel Monticeli^{1*}, Yasmine Mosleh², John-Alan Pascoe¹

CONTENTS



Model trained on 0//0, 0//45, 0//90 data; validated against 0//15 data

Modeling of manufacturing features

High fidelity FEA

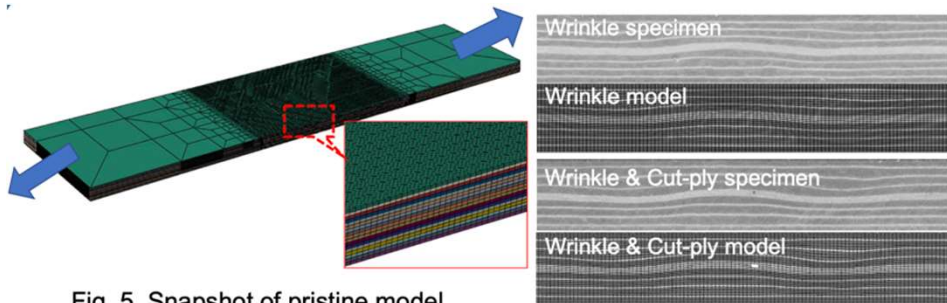


Fig. 5. Snapshot of pristine model.

Fig. 6. Cross sections of defect specimens and models.

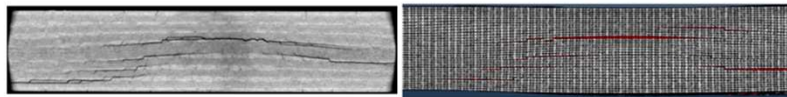


Fig. 7. CT observed and model predicted damage in the "wrinkle" case.

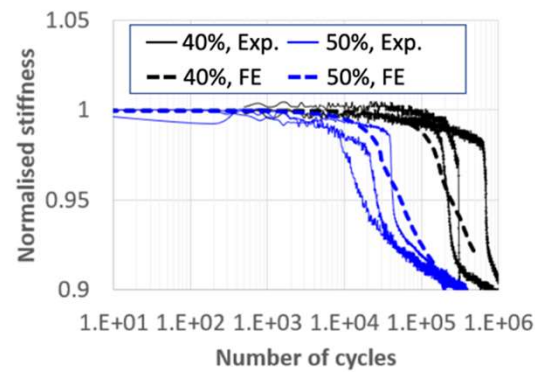


Fig. 8. Comparison between experiment and "wrinkle" model.

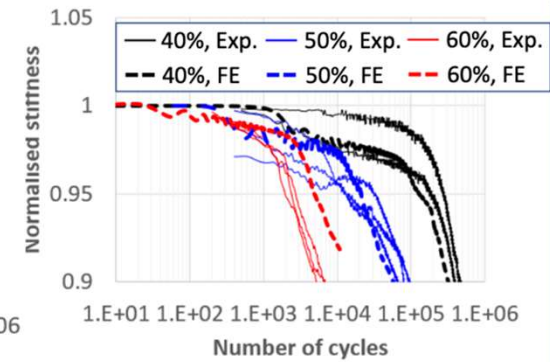
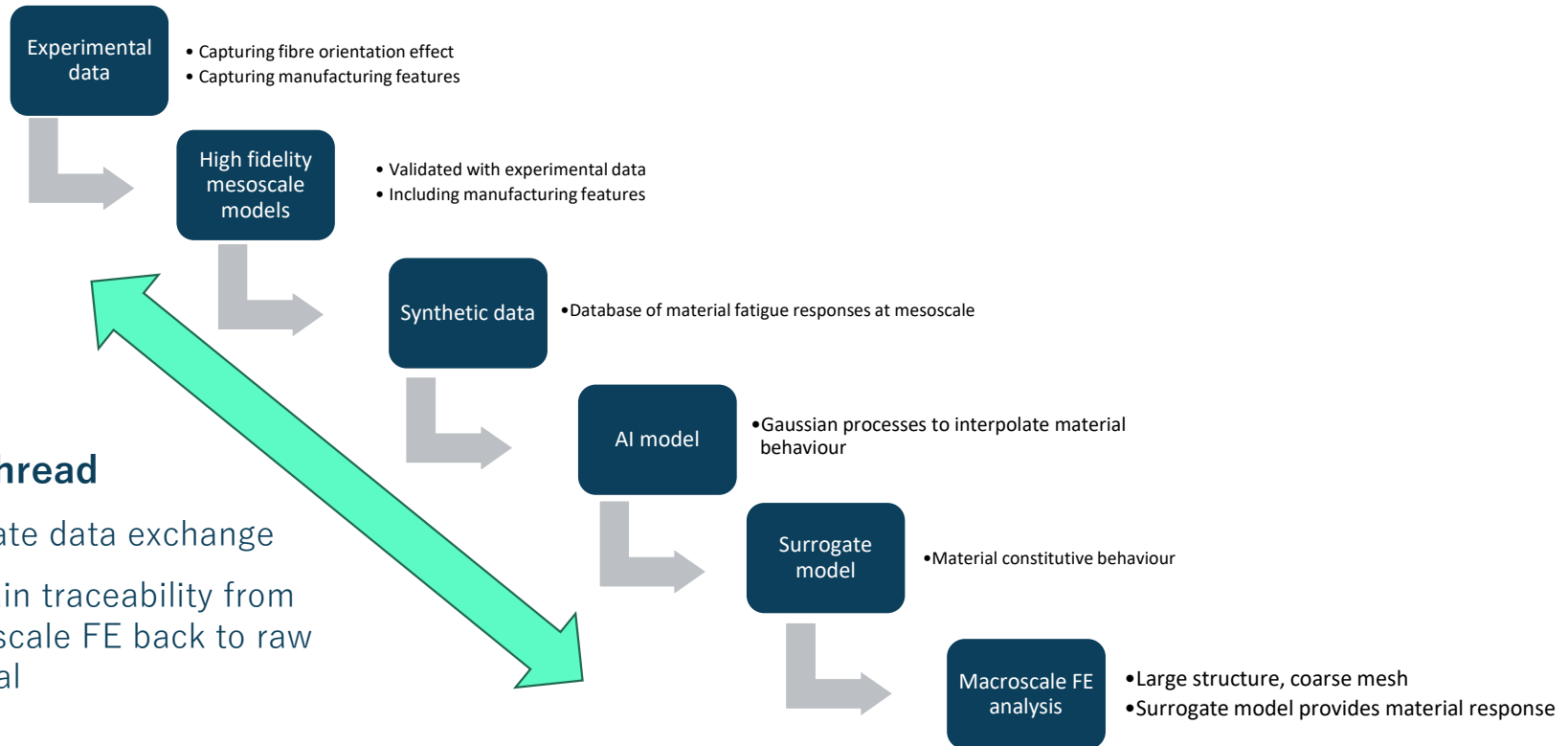


Fig. 9. Comparison between experiment and "wrinkle & cut-ply" model.

Bridging the scales

AI-based surrogate modelling

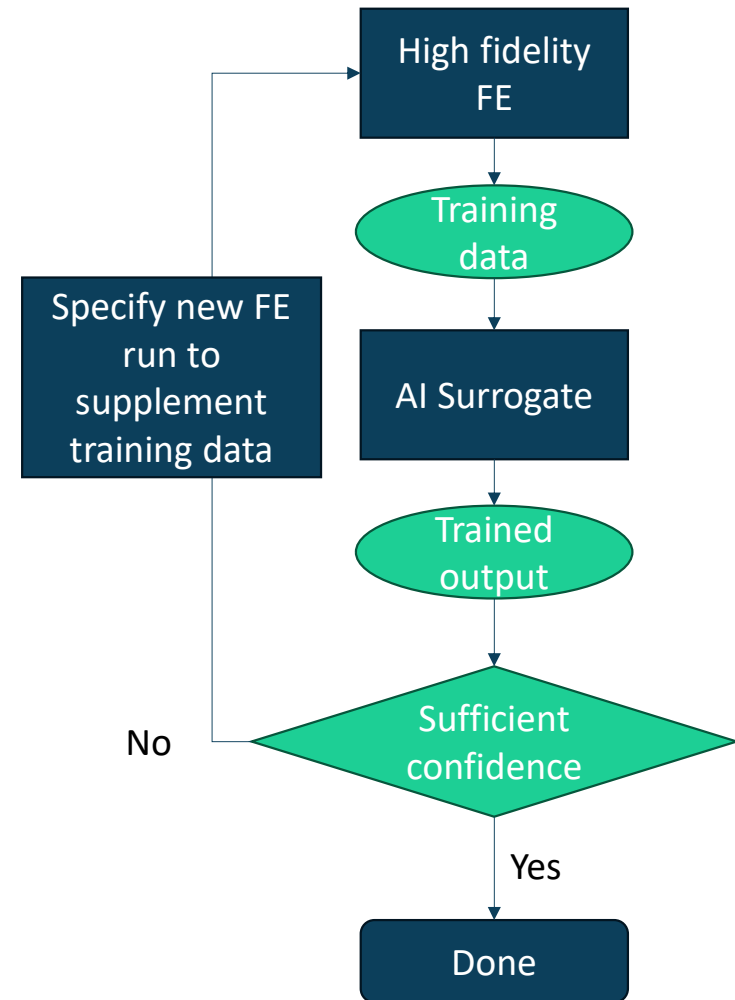


Digital thread

- › Facilitate data exchange
- › Maintain traceability from macroscale FE back to raw material

AI-based surrogate model

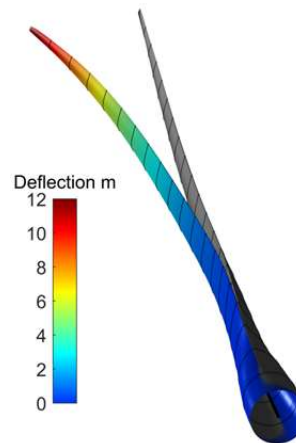
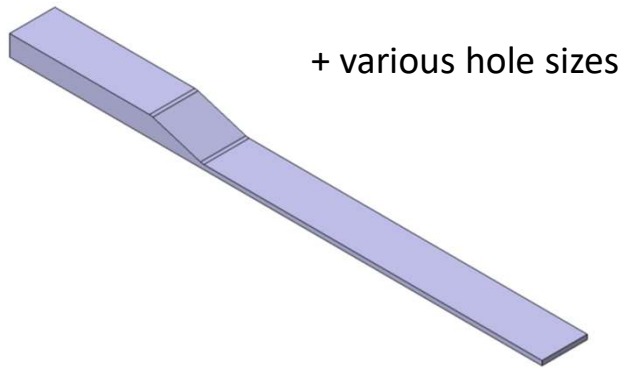
- › Gaussian process with adaptive sampling
- › Current status: training of quasi-static loading and fracture
- › **Inputs:** Load vector (3 normal components, 3 shear components). **Output:** Stress and strain tensors
- › Advantage of Gaussian process: information on certainty of model outcome, can 'order' new high-fidelity model runs to improve data.



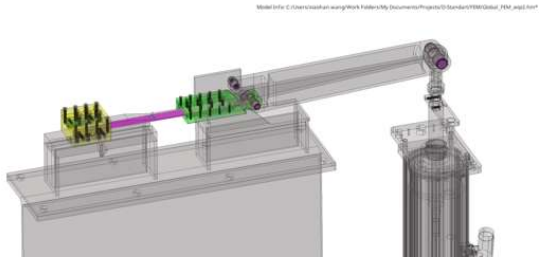
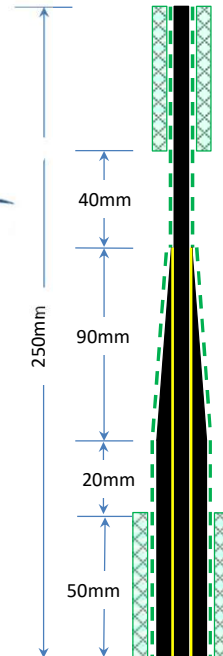
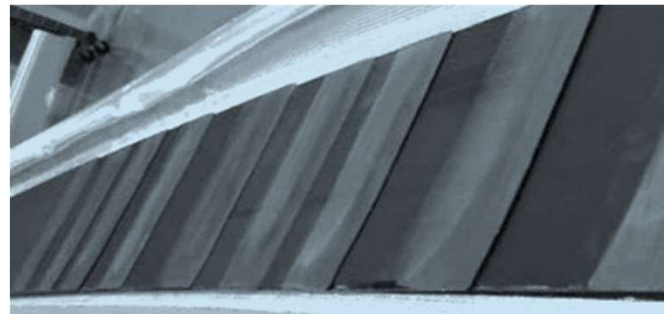
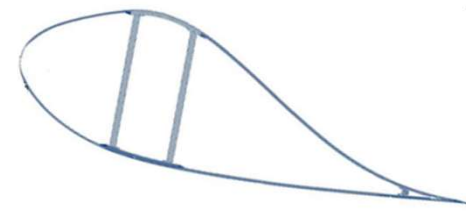
Validation cases

2 demonstrators

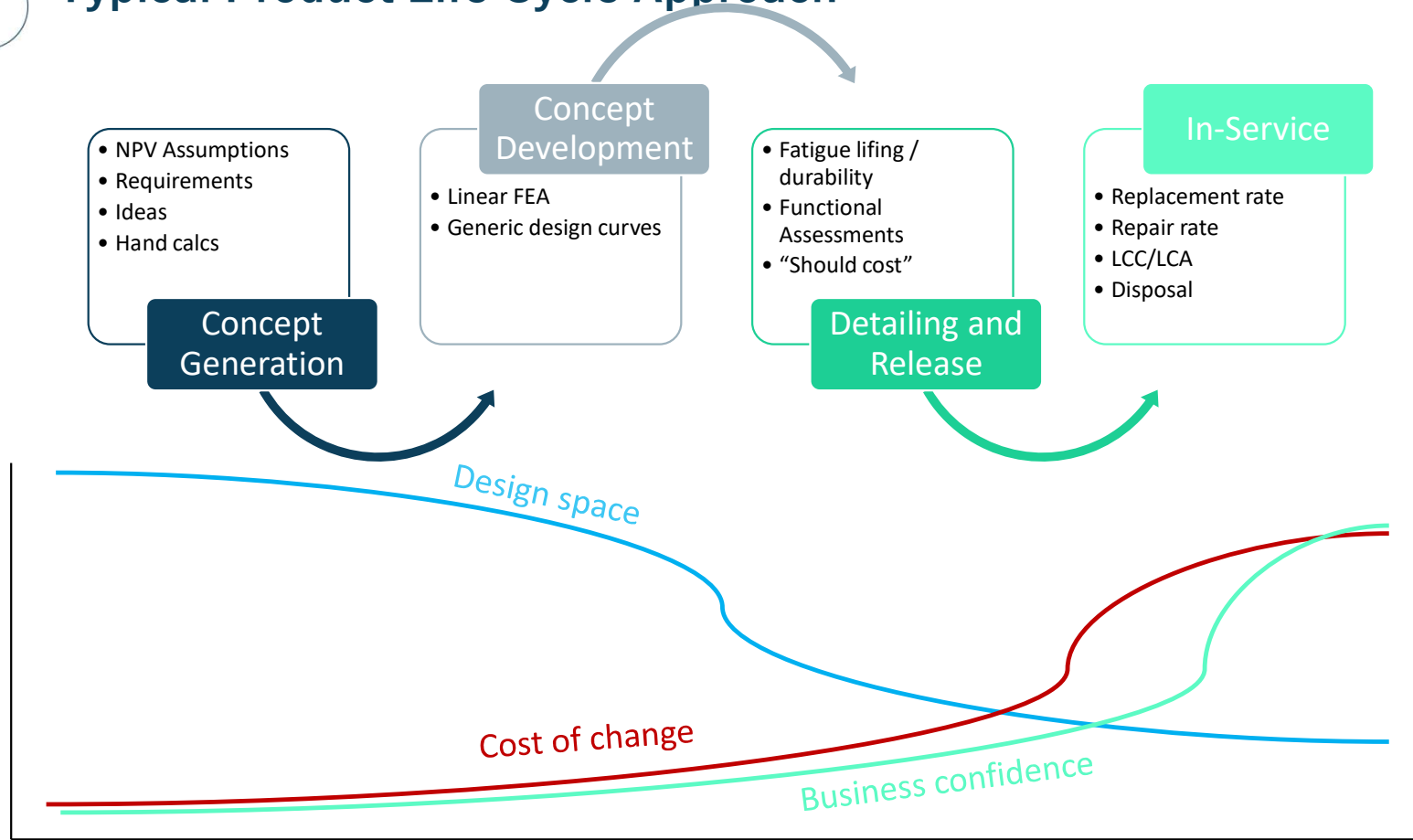
Aero demonstrator



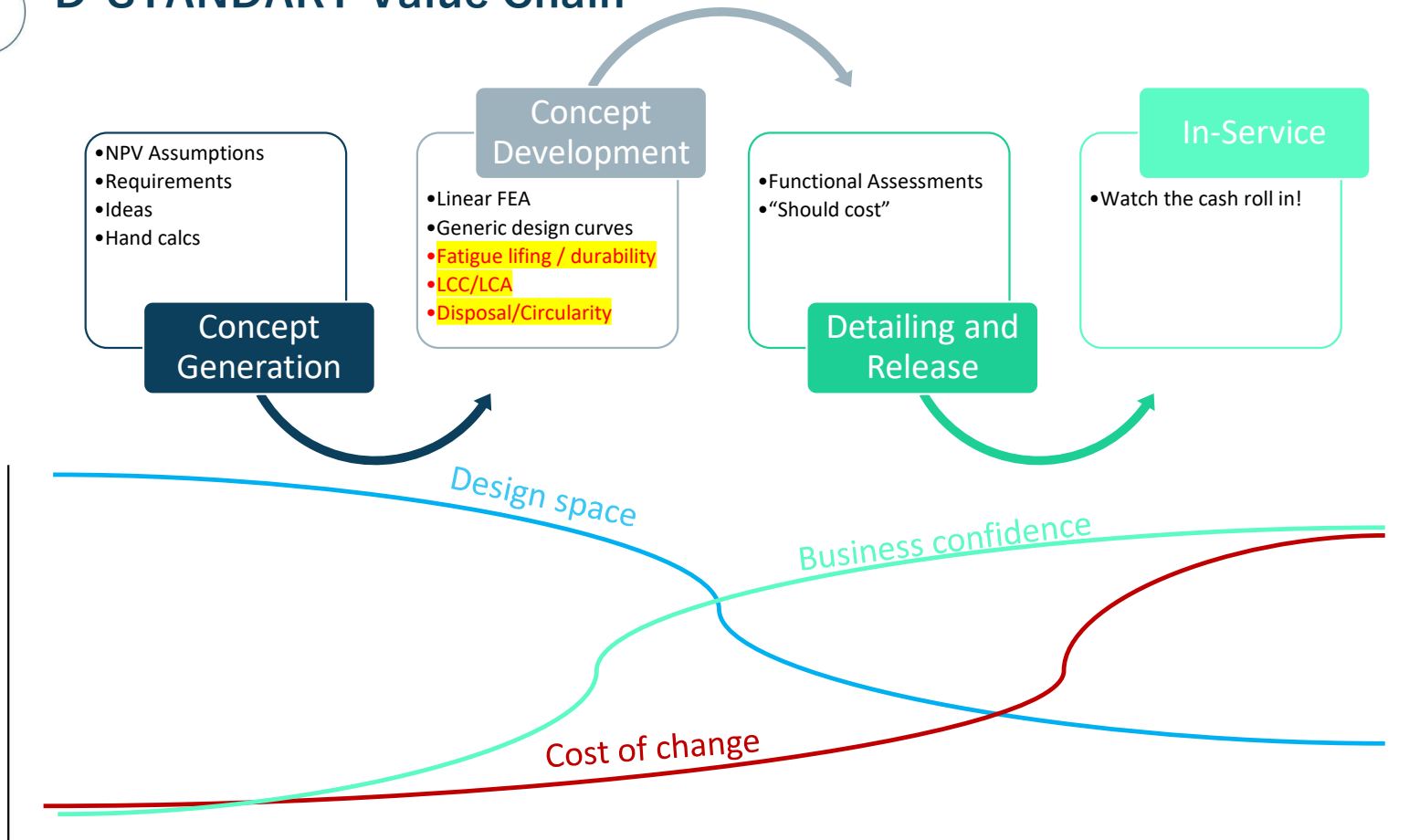
Wind demonstrator



Typical Product Life Cycle Approach



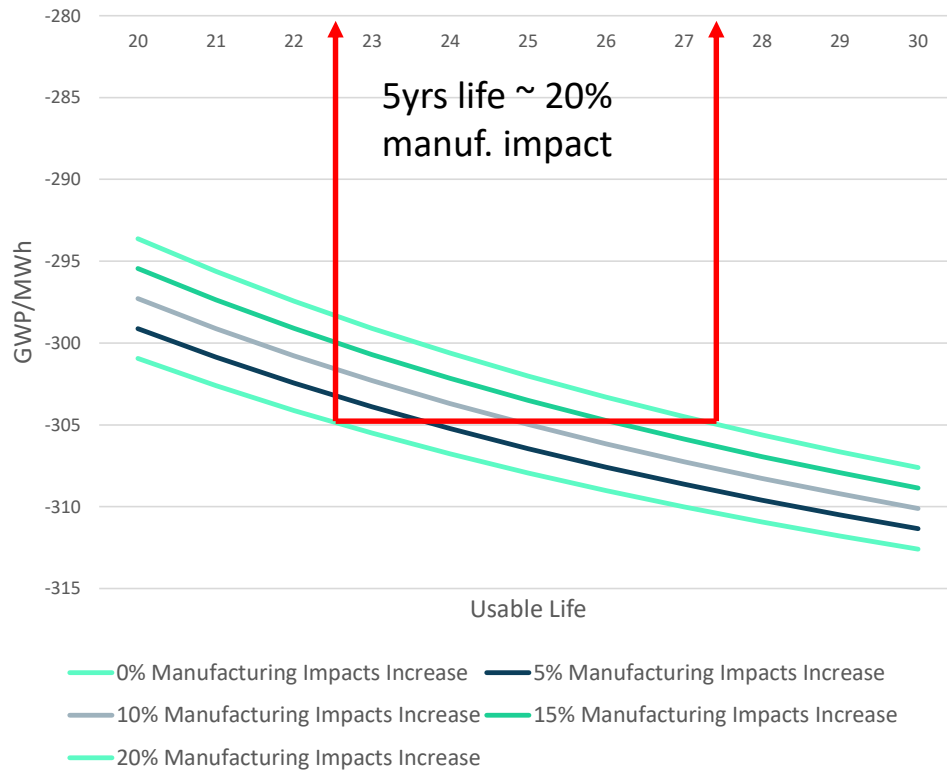
D-STANDART Value Chain



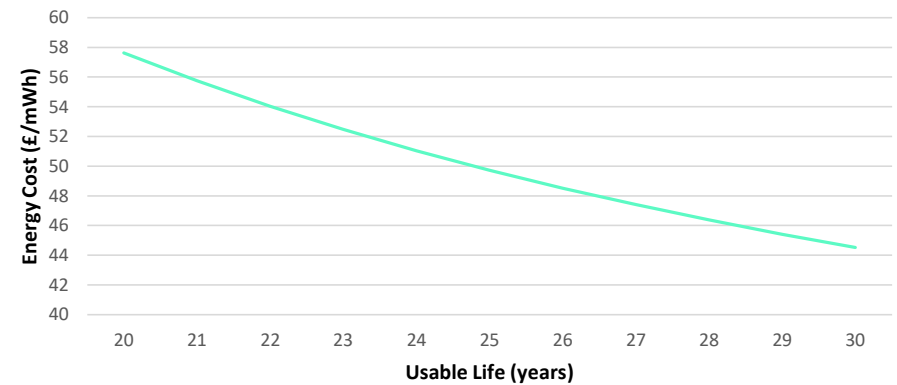
Relationship of LCA/C to Durability – Wind Energy Case Study

Wind Farm

Functional GWP with Life and Manufacturing Impact Increase



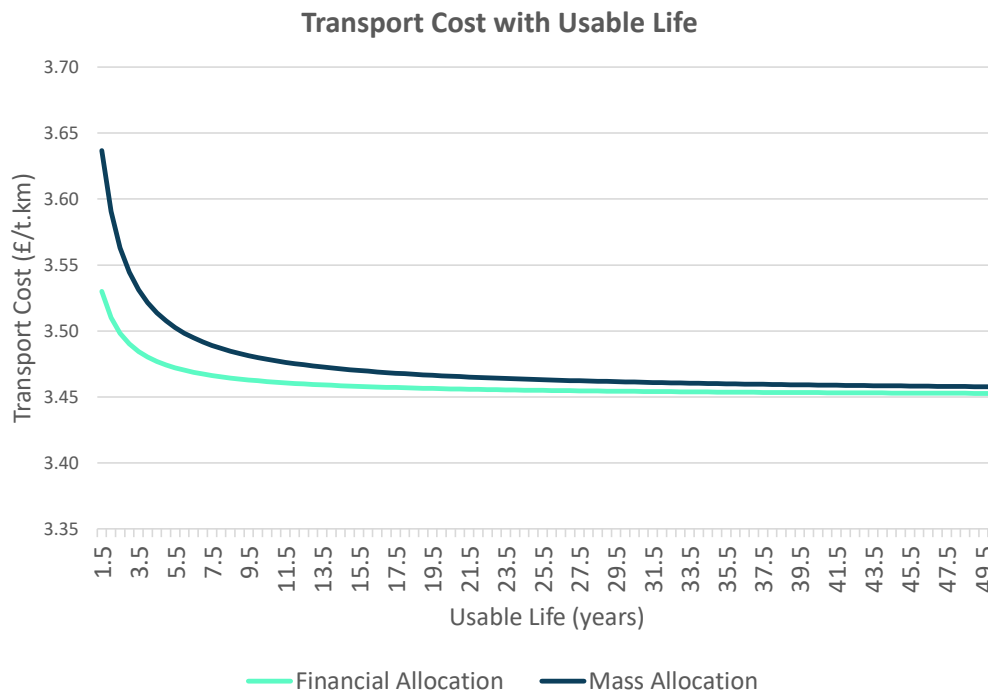
Energy Cost Reduction With Usable Life



Life Extension (years)	LCOE (\$/MWh)	Cost Savings (% from standard life)
(20+) 1	55.7	3.1
(20+) 5	49.7	13.9
(20+) 10	44.5	24.3

Relationship of LCA/C to Durability – Aerospace Case Study

Airbus A350



- › Manufacturing costs are quickly eclipsed by use phase, plateauing after 20 years
- › Extension of life has minimal cost reductions, with a 10-year life extension only reducing lifecycle cost by 0.07%

06

EMMC & EMCC

Alignment



EMMC & EMCC alignment

D-STANDART

Priority to align with EMMC and EMCC objectives at proposal stage

Five complimentary activities:



Test methodology



Material characterisation



AI-enable material modelling



Life-cycle assessment



Digital thread

Yielding five distinct and tangible outcomes:

- › Approach to accelerated fatigue test standard
- › A repository of material data under fatigue loading
- › A repository of multi-scale models capable of modelling “defect-free” layups and specimens with manufacturing defects
- › Demonstrating the impact of our approach through TEA and LCA
- › A roadmap aligned with EMCC/EMMC to include proposed ongoing research and development prospects

EMMC & EMCC contributions

Addressing their objectives

Contribution to EMMC objectives

- › “to promote modelling by means of physics-based and data-driven models in industry” – EMMC White Paper
- › “The goal is to strengthen the link between materials modelling and experiments by developing improved post-processing models with the necessary physics contents [...]” – EMMC website

Contribution to EMCC objectives

- › “To support establishing a community of European stakeholders in the process of developing and improving characterisation tools in order to bring [...] advanced materials in Europe into end products more successfully.” – EMCC website
- › “[...] roadmap for characterisation techniques for engineering and upscaling of [...] advanced materials in Europe. [...] support the strengthening of Europe’s industrial capacity and competitiveness” – EMCC website



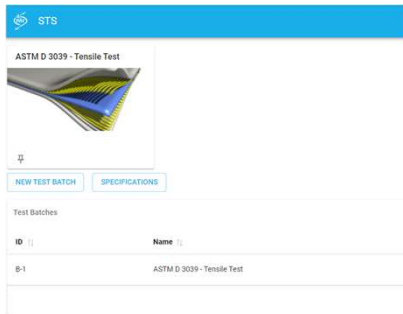
D-STANDART specific objective: **Develop fatigue behaviour models** at the **micro-and meso-scale** levels, and **AI surrogate models** at the **meso-scale** level. To develop a finite-element based, validated probabilistic meso-mechanical modelling framework for the prediction of the fatigue performance of laminates. Exploit AI trained surrogate models for **fatigue prediction of macro-scale structures**.

Chapter 6

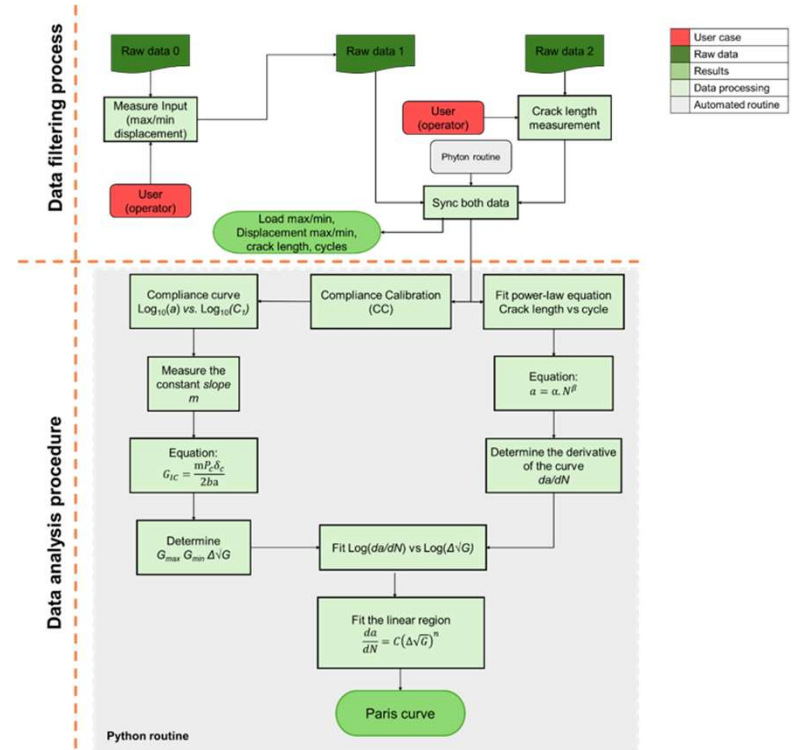
EMMC & EMCC contributions

Use of CHADA/MODA/EMMO

- CHADA and MODA templates in use for meta-data capture within D-STANDART
- CHADA / EMMO integrated into digital thread infrastructure



1. Plate		
1.1	Number	Group 1
1.2	Name	TUD-Group 1 - (β-θ) - 01
1.3	Name convention	Figure 2 TUD - Institution Group 1 = The first manufacturing group (each group has 4 plates) (β-θ) = Orientation of the delamination layer (0/90) 01 = Number of repetition of the same configuration M37852 Unidirectional carbon fiber composite (Hexcel)
1.4	Material	
1.5	Lay-up	(0 ₂ /90 ₂) ₂
1.6	Organisation	Note: Double slash (//) means the delamination region Technische Universiteit Delft (TUD)
1.7	User	Francisco Mocal Montiel
1.8	Manufacturer data sheet	Document reference: Figure 2 heavily datasheet Hand lay-up to manufacture the laminate plates. - Process type: Autoclave - Cure step 1: 110°C for 60 min - Cure step 2: 180°C for 120 min - Pressure: 7 bar - Vacuum: 0.2 bar



07


Cross-fertilisation

Synergy with sister projects



Cross-fertilisation

Synergy with sister projects



Project No.	Acronym	Title
101091409	D-STANDART	Durability Modelling of Composite Structures with Arbitrary Lay-up using Standardized Testing and Artificial Intelligence
101091621	AddMorePower	Advanced modelling and characterization for power semiconductor materials and technologies
101091534	KNOWSKITE-X	Knowledge-driven fine-tuning of perovskite-based electrode materials for reversible Chemicals-to-Power devices
101091687	MatCHMaker	Open data and industry driven environment for multiphase and multiscale Materials Characterization and Modelling combining physics and data-based approaches
101092211	CoBRAIN	Integrated Computational-Experimental material Engineering of Thermal Spray coatings
101091912	AID4GREENEST	AI powered characterization and modelling for GREEN STEEL technology

Thank you!

Contact points for any question:

FOLLOW US ON:



zenodo

d-standart.eu



› Coordinator

- › Marco NAWIJN (NLR)
- › Email: Marco.Nawijn@nlr.nl
- › Phone: +31 88 511 44 86

› Project Management Officer/ Communication

- › Peggy FAVIER (LUP)
- › Email: peggy.favier@l-up.com
- › Phone: +33 67 56 41 07 8



Funded by
the European Union

Funded by the European Union. Views and opinions expressed are however those of the author(s) only and do not necessarily reflect those of the European Union. Neither the European Union nor the granting authority can be held responsible for them.



UK Research
and Innovation

