



# Materials & Structural Health Monitoring

## Engineering Mechanics Symposium – Workshop Introduction

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26<sup>th</sup> of October 2022

- 1 Introduction
- 2 What Structural Health Monitoring is
- 3 Why Structural Health Monitoring is needed
- 4 Disciplines of Structural Health Monitoring
- 5 Workshop Presentations
- 6 Bibliography

## Organisers of the Workshop “Materials & Structural Health Monitoring”



University of Twente



Delft University of  
Technology



Eindhoven University of  
Technology

### Wikipedia

Structural health monitoring (SHM) involves the observation and analysis of a system over time using periodically sampled response measurements to monitor changes to the material and geometric properties of engineering structures such as bridges and buildings.

### The Constructor – Building Ideas

Structural health monitoring (SHM) is the process of using damage detection and characterization techniques for critical structures like bridges, wind turbines, and tunnels. It is a non-destructive in-situ structural evaluation method that employs several types of sensors embedded or attached to the structure.

## IGI-Global

- 1 Systematic procedure used to evaluate the damage of engineering structures such as bridges.
- 2 The integration of sensors and actuators in a material or structure in order to perform load monitoring, damage detection, damage diagnosis and damage prognosis so that nondestructive testing becomes online and in situ.
- 3 Process of implementing a damage detection and characterization strategy for engineering structures.
- 4 Structural Health Monitoring (SHM) is the process of detecting and characterizing damages in engineering structures to ensure structural integrity and safety of buildings, bridges and all kinds of engineering structures. The monitoring is automated and can continuously monitor and detect state (such as strain, stress, and temperature) and damages (such as cracks, holes, and ruptures) of structures with minimum human intervention. Using SHM we can optimize the maintenance of the structures maximizing the service life of the structure and the safety of its usage.

## Summarizing, Structural Health Monitoring (SHM):

- is process / method with several aspects:
  - Data acquisition through embedded sensor systems
  - Processing of acquired data
  - Analysis of processed data
- is applied to various fields of engineering
- focusses on detection and characterization (optional) of damage

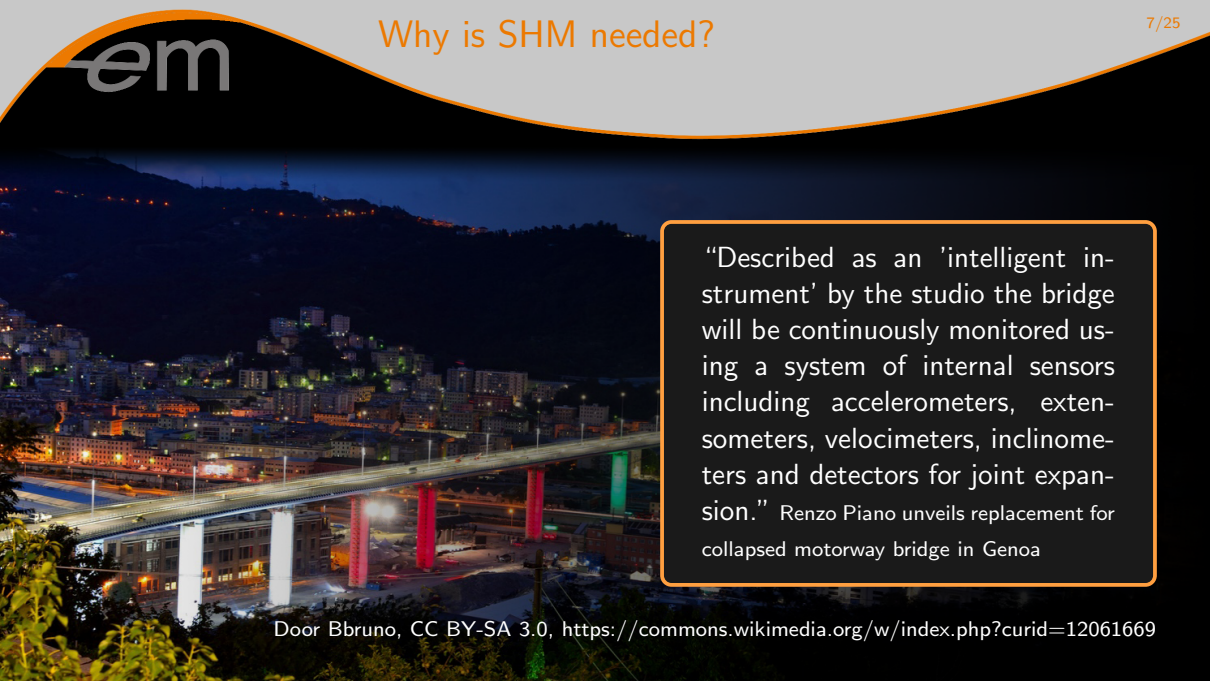
### Ponte Morandi

- Built between 1963 and 1967
- Collapsed in 2018
- Far from expected end of life
- Lack of maintenance identified as cause
- 43 casualties

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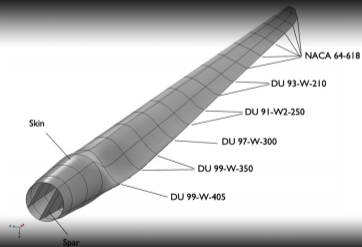
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“Described as an ‘intelligent instrument’ by the studio the bridge will be continuously monitored using a system of internal sensors including accelerometers, extensometers, velocimeters, inclinometers and detectors for joint expansion.” Renzo Piano unveils replacement for collapsed motorway bridge in Genoa

## Design Process



Source: Comsol

## Manufacturing Process



Source: LM Wind

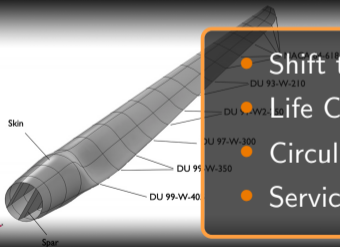
## Operation



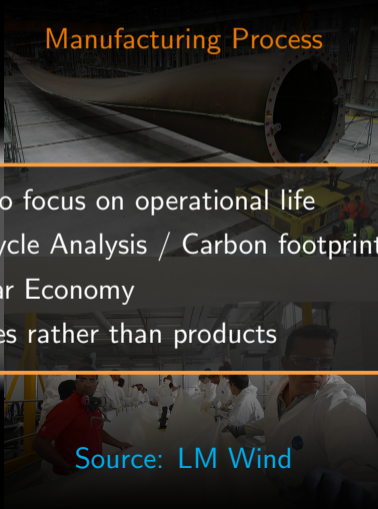
## Design Process

## Manufacturing Process

## Operation



- Shift to focus on operational life
- Life Cycle Analysis / Carbon footprint
- Circular Economy
- Services rather than products



Source: Comsol

Source: LM Wind

Summarizing, Structural Health Monitoring is needed to guarantee safe operation of systems throughout their operational life, while:

- Accounting for changes and variability in operational conditions
- Designing less conservative, use less material, optimizing performance, reducing CO2 emissions
- Optimizing maintenance planning, avoiding unnecessary waste of materials

- I All materials have inherent flaws or defects
- II The assessment of damage requires a comparison between two systems states
- III Identifying the existence and location of damage can be done in an unsupervised learning mode, but identifying the type of damage and the damage severity can generally only be done in a supervised learning mode
- IV
  - a Sensors cannot measure damage. Feature extraction through signal processing and statistical classification is necessary to convert sensor data into damage information
  - b Without intelligent feature extraction, the more sensitive a measurement is to damage, the more sensitive it is to changing operational and environmental conditions
- V The length- and time-scales associated with damage initiation and evolution dictate the required properties of the SHM sensing system
- VI There is a trade-off between the sensitivity to damage of an algorithm and its noise rejection capability
- VII The size of damage that can be detected from changes in system dynamics is inversely proportional to the frequency range of excitation

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- Development new sensor technologies: PZT, optical fibres, MEMS, ...
- Autonomous, distributed, wireless sensor networks: local processing, data transfer, power management, ...

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- Material Science: Wear, fatigue, fracture, degradation, corrosion, ...

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## Data Analysis

- Statistical pattern recognition
- Machine Learning & Artificial Intelligence



## Digital Twins & Cyber Physical Spaces

- Virtual representation of physical objects and/or behaviour
- One-way or two-way interaction between virtual and physical space
- Sensor & Actuation technologies for interaction
- Scenario projection

**UTwente**

- PZT-based guided waves
- Nonlinear ultrasound
- Optical fibres
- Vision based inspection
- Autonomous distributed wireless sensor networks

**TU Delft**

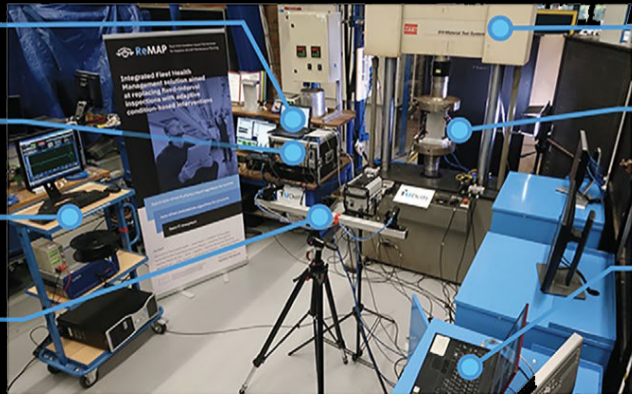
- Near-NDT concrete monitoring
- Magneto-based contact-free sensors
- (Active &) Passive ultrasound
- Optical Fibres
- Self sensing of impact
- Sensor fusion

Fiber Bragg gratings

Lamb waves

Distributed strain sensing

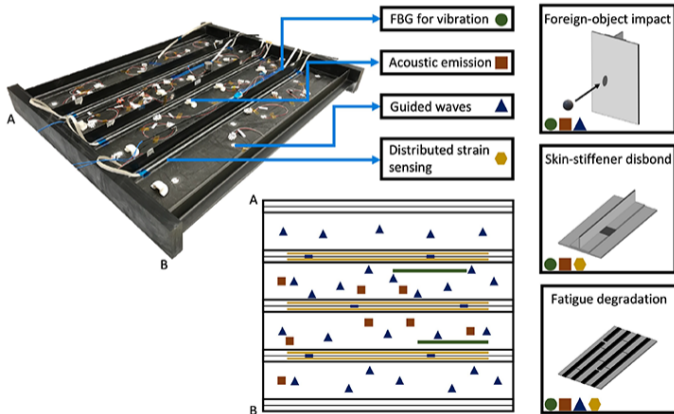
Digital image correlation



MTS machine

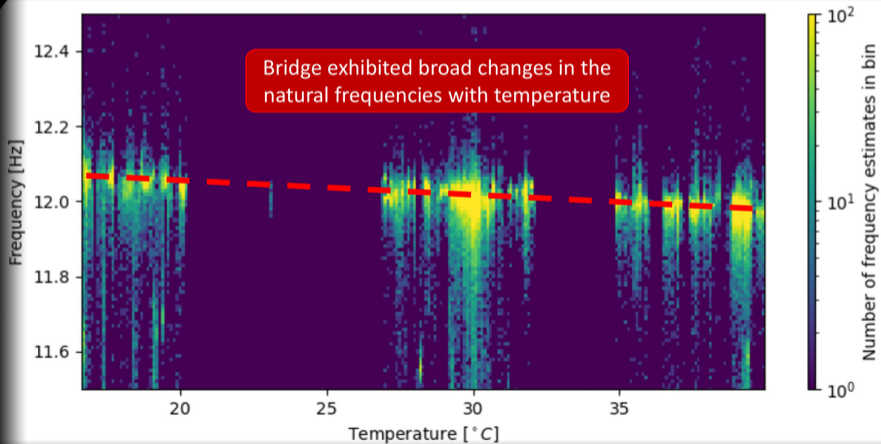
Stiffened composite panel

Acoustic emission







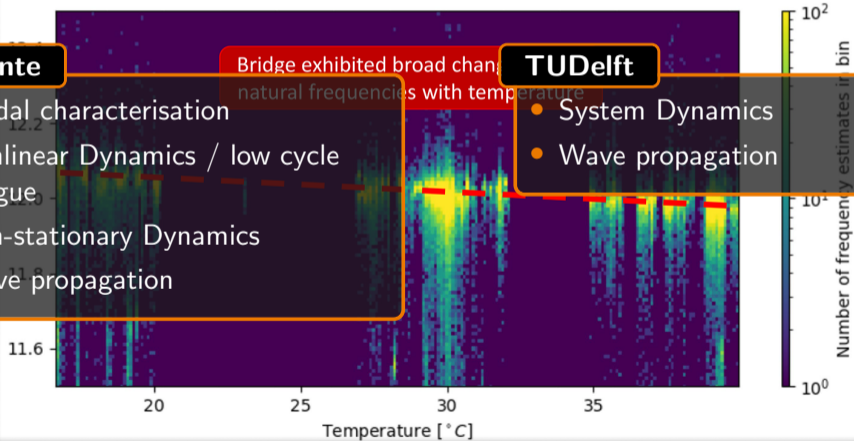


**UTwente**

- Modal characterisation
- Nonlinear Dynamics / low cycle fatigue
- Non-stationary Dynamics
- Wave propagation

**TU Delft**

- System Dynamics
- Wave propagation



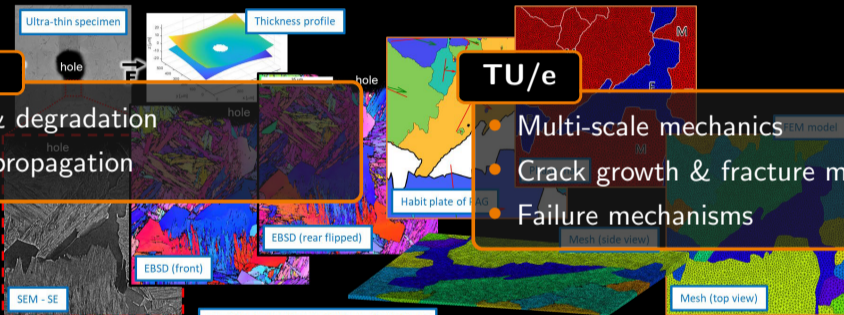


## UTwente

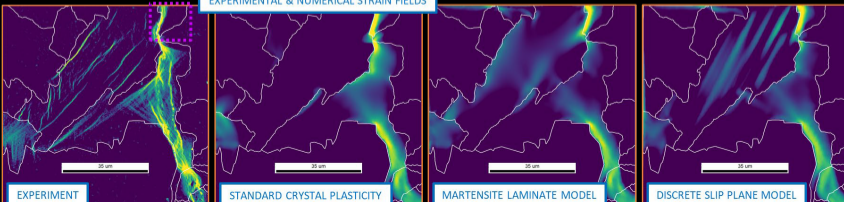
- Wear & degradation
- Crack propagation

## TU/e

- Multi-scale mechanics
- Crack growth & fracture mechanics
- Failure mechanisms



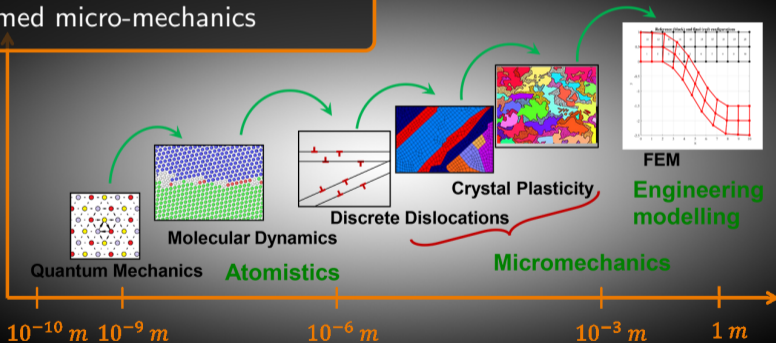
### EXPERIMENTAL & NUMERICAL STRAIN FIELDS

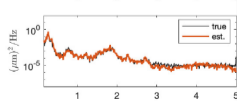
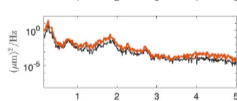
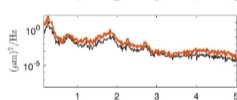
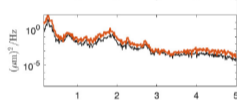
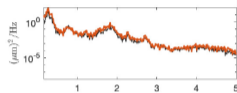
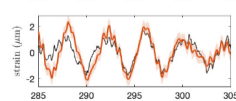
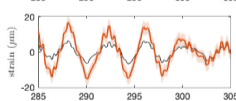
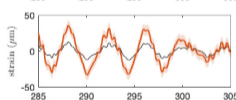
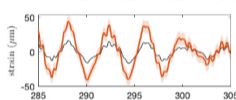
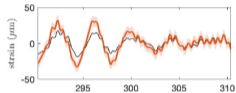
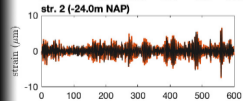
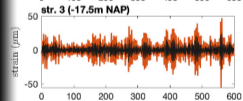
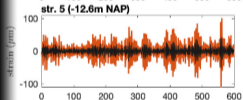
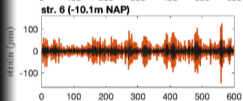
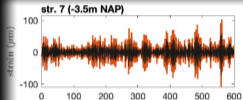
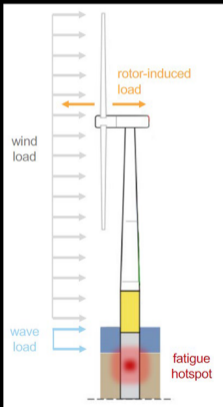
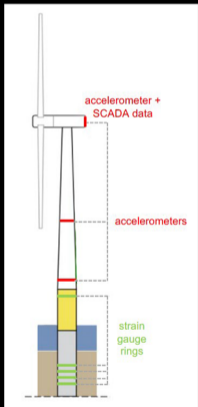


RUG

- Atomistic-informed micro-mechanics

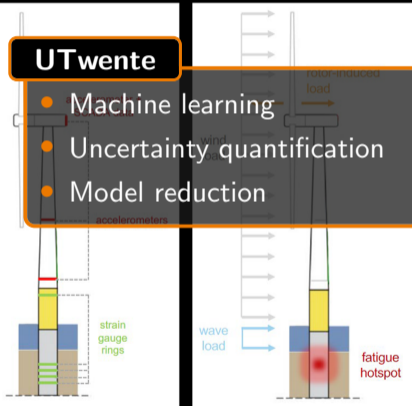
time ↑





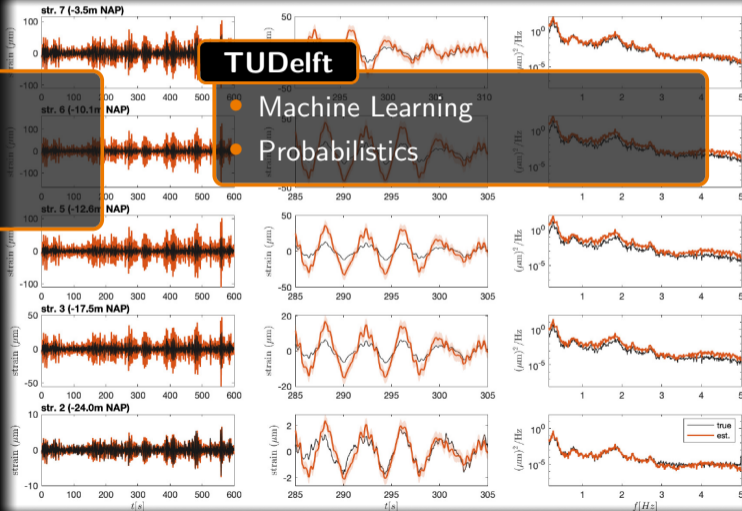
## UTwente

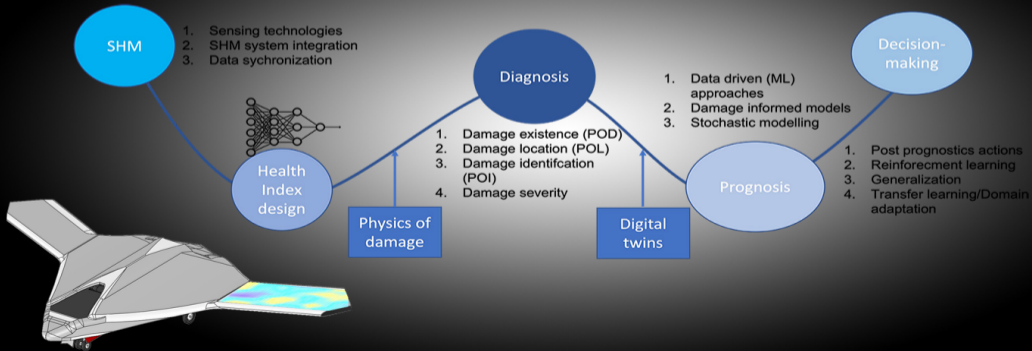
- Machine learning
- Uncertainty quantification
- Model reduction



## TU Delft

- Machine Learning
- Probabilistics





## UTwente

- Non-stationary dynamics

SHM

1. Sensing technologies
2. SHM system integration
3. Data synchronization

Health  
Index  
designPhysics of  
damage

Diagnosis

1. Damage existence (POD)
2. Damage location (POL)
3. Damage identification (POI)
4. Damage severity

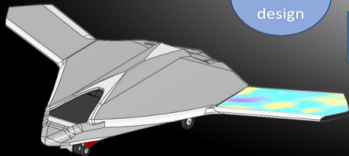
Digital  
twins

Prognosis

1. Data driven (ML) approaches
2. Damage informed models
3. Stochastic modelling

Decision-  
making

1. Post prognostics actions
2. Reinforcement learning
3. Generalization
4. Transfer learning/Domain adaptation



Naval

Railway

- Infrastructure
- Rolling Stock

Large infrastructure

- Static
- Moveable

Buildings

Wind Energy

Process Industry

Aerospace

Rotating Systems

- Engines
- Bearings

Water Distribution

- Drink water
- Waste water

Smart Industry

Naval

Railway

- Infrastructure
- Rolling Stock

Large infrastructure

- Static
- Moveable

Buildings

Wind Energy

<sup>NEW</sup> Energy Systems & Storage

Process Industry

Aerospace

Rotating Systems

- Engines
- Bearings

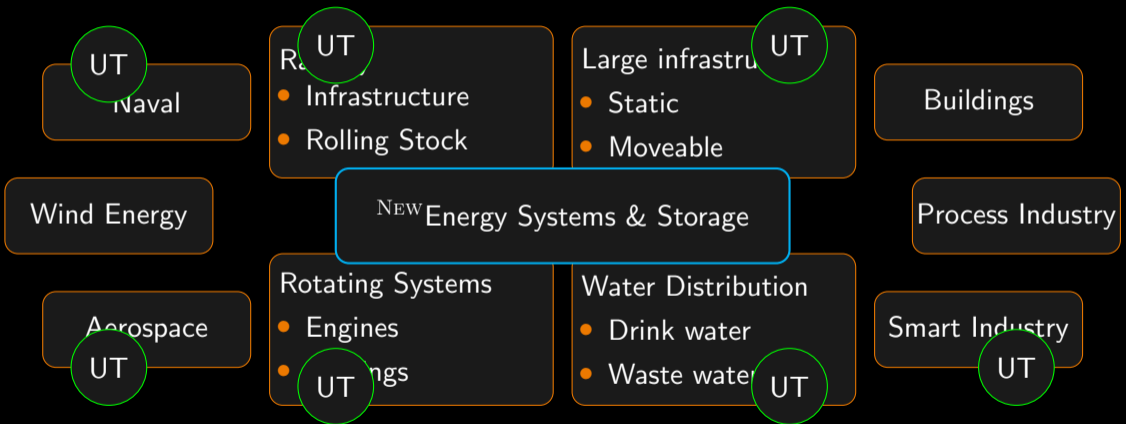
Water Distribution

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- Waste water

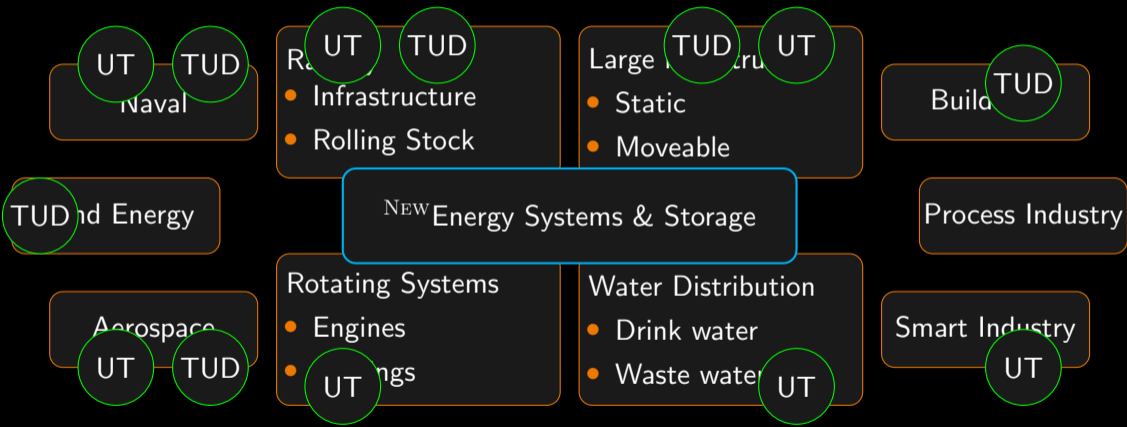
Smart Industry



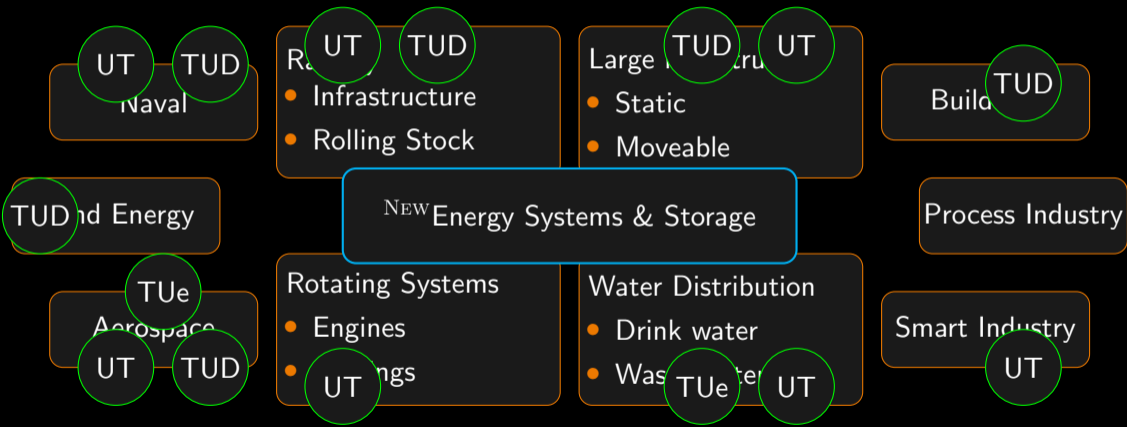
# SHM Application Fields



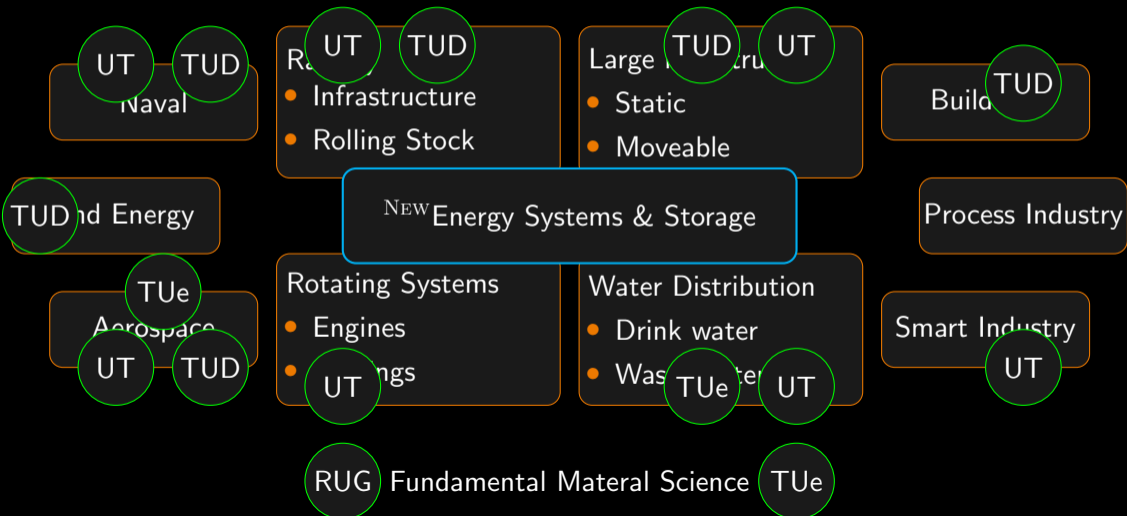
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- Fusion of sensors and data sources



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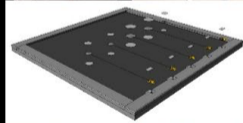
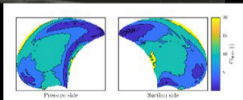
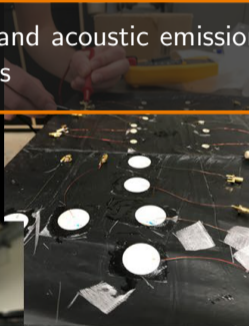
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Summarizing, main challenges are found in:

- Complex, non-isotropic and/or nonlinear material behaviour
- Integration of sensors & sensor technologies
- Fusion of sensors and data sources
- Management of data and conversion to information
- Providing meaningful information in a meaningful way to end-users

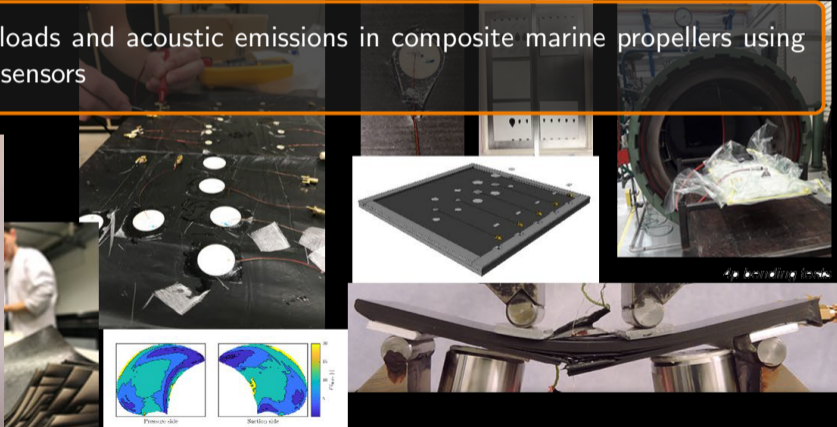
**Arno Huijer**

Monitoring of dynamic loads and acoustic emissions in composite marine propellers using embedded piezoelectric sensors



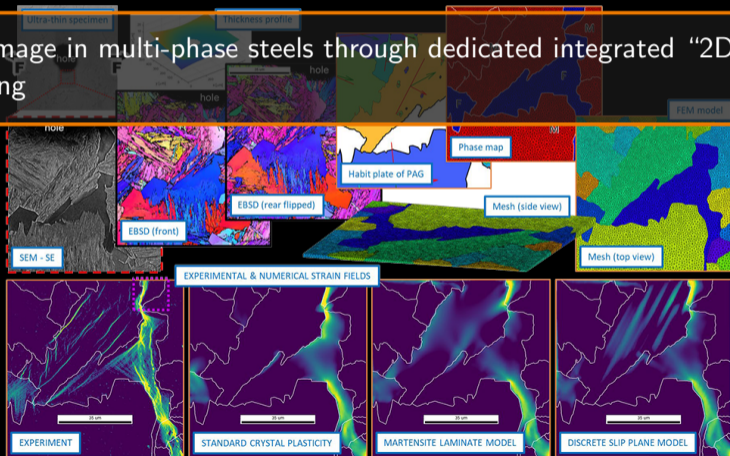
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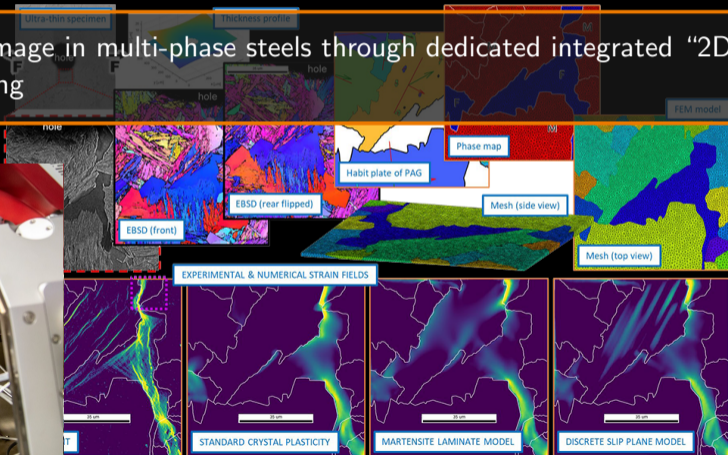
**Tijmen Vermeij**

Unraveling plasticity and damage in multi-phase steels through dedicated integrated “2D” experimental-numerical testing



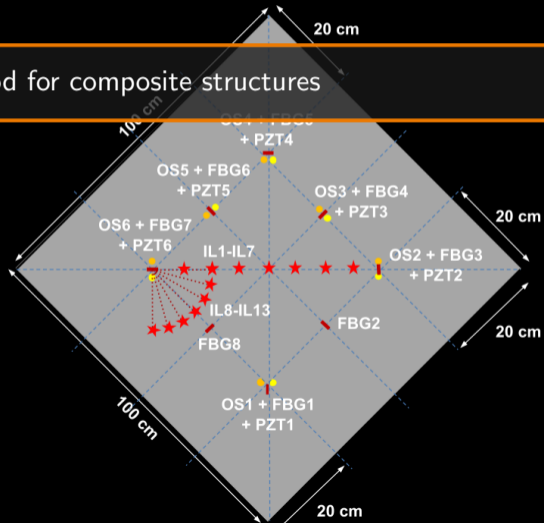
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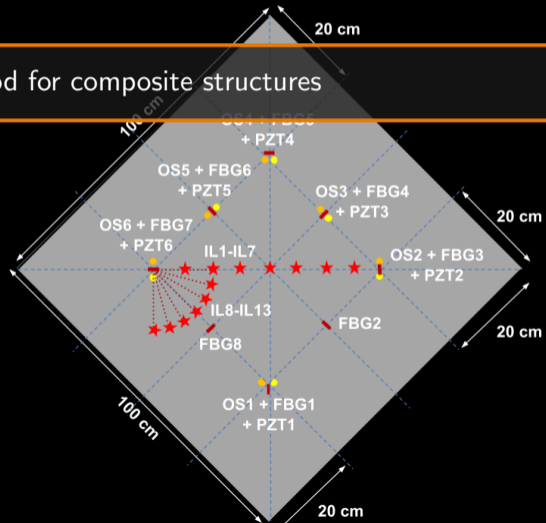
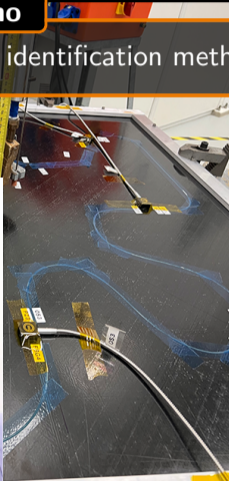
## Natália Ribeiro Marinho

Dynamics-based impact identification method for composite structures



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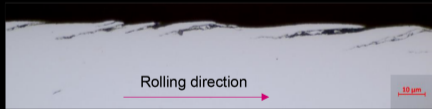
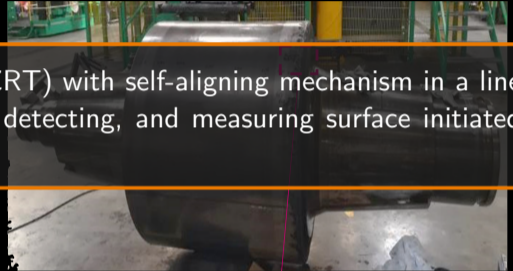
Dynamics-based impact identification method for composite structures





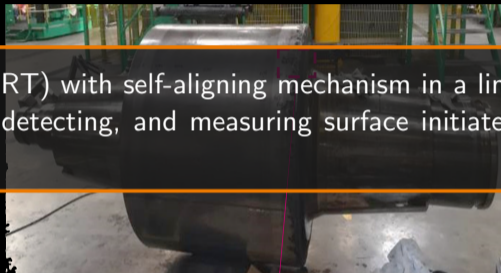
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Newly developed Cam-Roller follower Tester (CRT) with self-aligning mechanism in a line contact configuration and ways of simulating, detecting, and measuring surface initiated RCF cracks



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- [1] Wikipedia, *Structural Health Monitoring*, [https://en.wikipedia.org/wiki/Structural\\_health\\_monitoring](https://en.wikipedia.org/wiki/Structural_health_monitoring)
- [2] The Constructor – Building Ideas, *What is Structural Health Monitoring in Civil Engineering?*, <https://theconstructor.org/digital-construction/structural-health-monitoring-civil-engineering/554160/>
- [3] IGI-Global, *What is Structural Health Monitoring*, <https://www.igi-global.com/dictionary/structural-health-monitoring/42149>
- [4] LM Wind, LM Wind Power manufactures the world's first wind turbine blade beyond 100 meters!, [www.lmwindpower.com/.../lm-wind-power-manufactures-the-worlds-first-wind-turbine-blade-beyond-100-meters](http://www.lmwindpower.com/.../lm-wind-power-manufactures-the-worlds-first-wind-turbine-blade-beyond-100-meters)
- [5] Comsol, *Anayising Wind Turbine Blades with the Composite Materials Module*, [www.comsol.com/blogs/analyzing-wind-turbine-blades-with-the-composite-materials-module/](http://www.comsol.com/blogs/analyzing-wind-turbine-blades-with-the-composite-materials-module/)
- [6] K. Worden, C.R. Farrar, G. Manson, G. Park, *The fundamental axioms of structural health monitoring*, *Proceedings of the Royal Society A* (2007) **463**, 1639-1664



# Materials & Structural Health Monitoring

## Engineering Mechanics Symposium – Workshop Introduction

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