

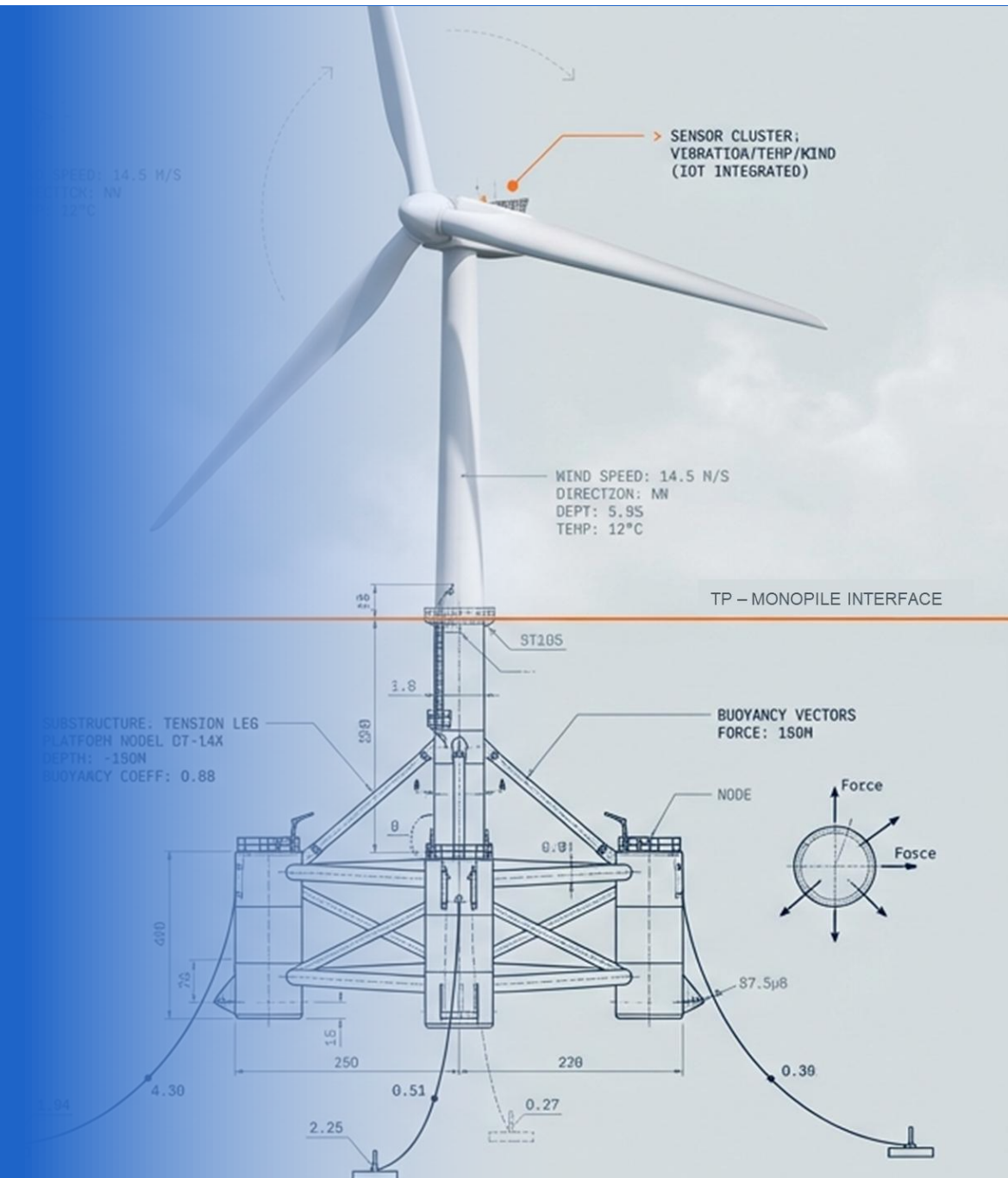


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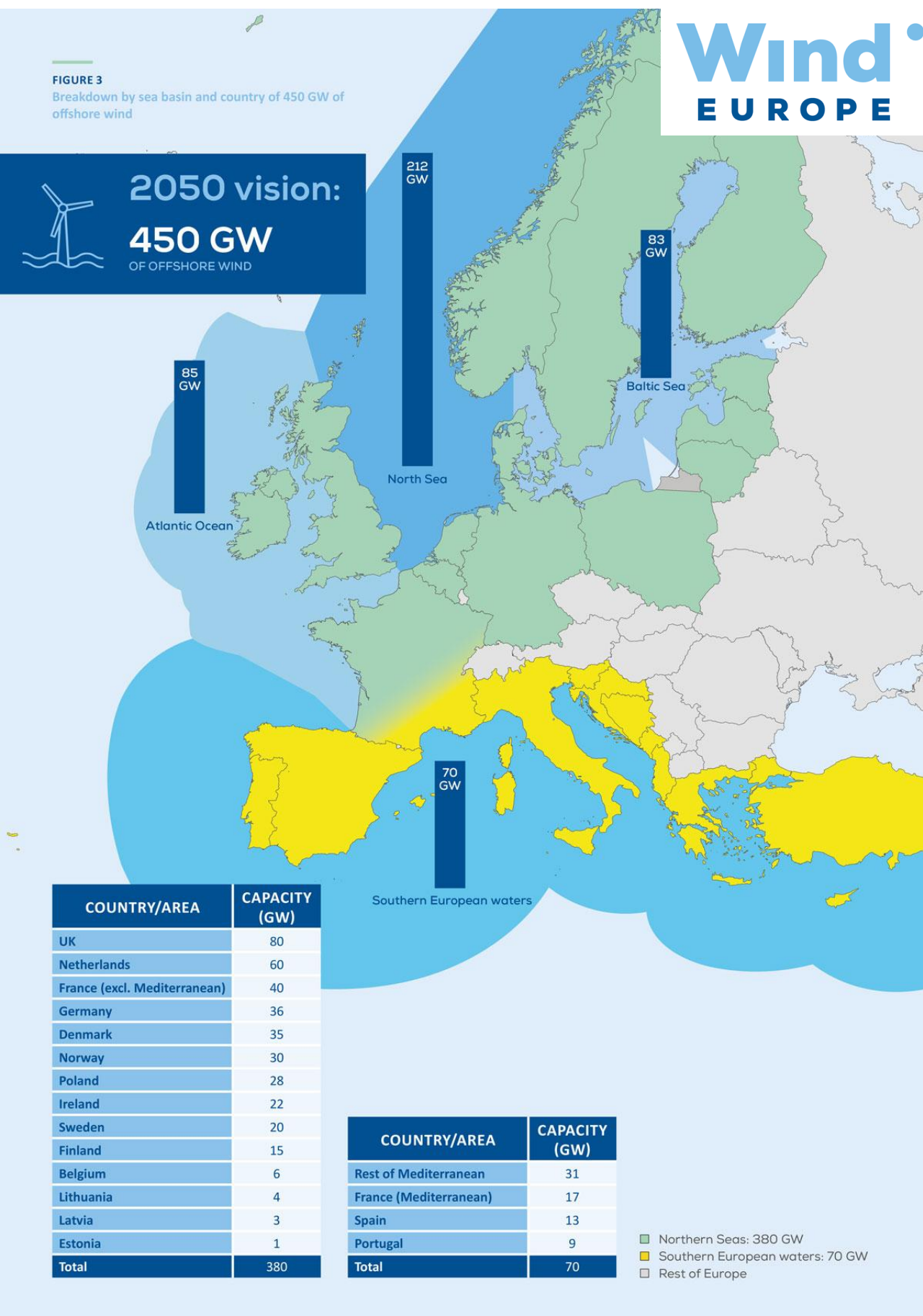
Advancements in digital twin frameworks for offshore wind: ongoing research and future directions

Prof. dr. ing. Kris Hectors

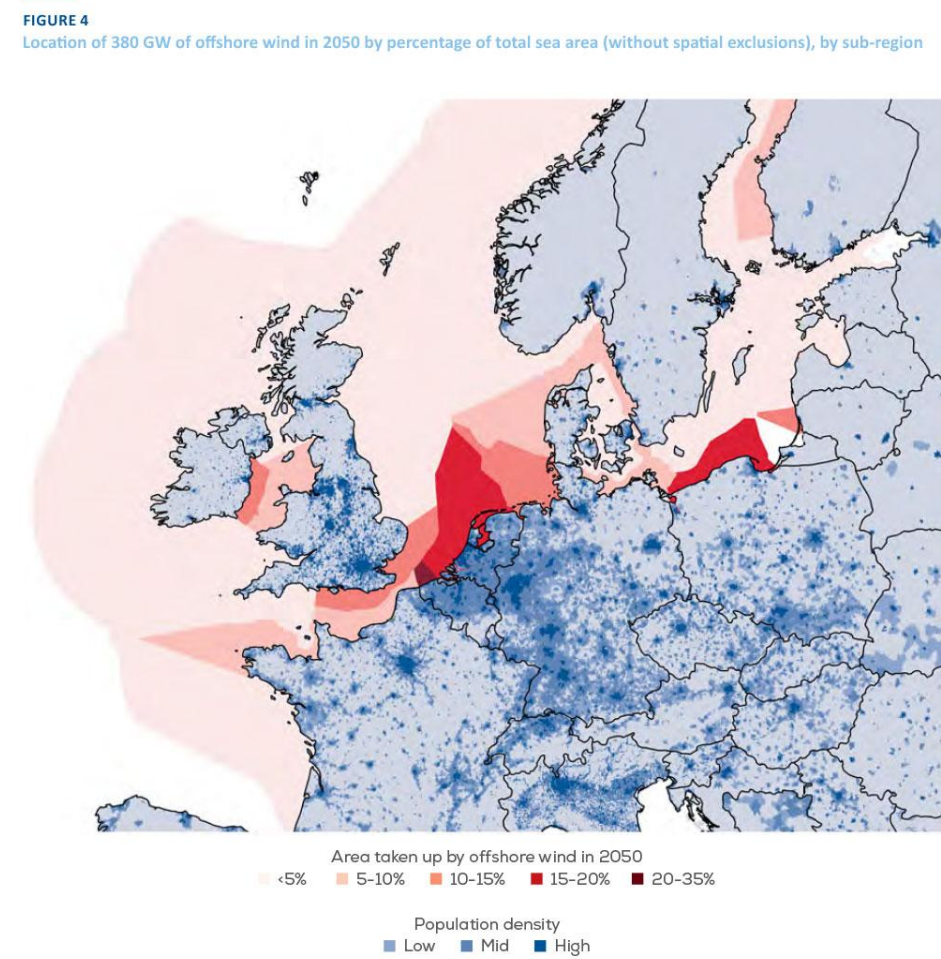
25/03/2026



Goals of offshore renewable generation in 2050



380 GW of offshore wind in the North Seas would require less than 3% of the total area of the North Seas.



European union:



2030

86-89 GW

2040

259-261 GW

2050

356-366 GW

OFFSHORE WIND FARM

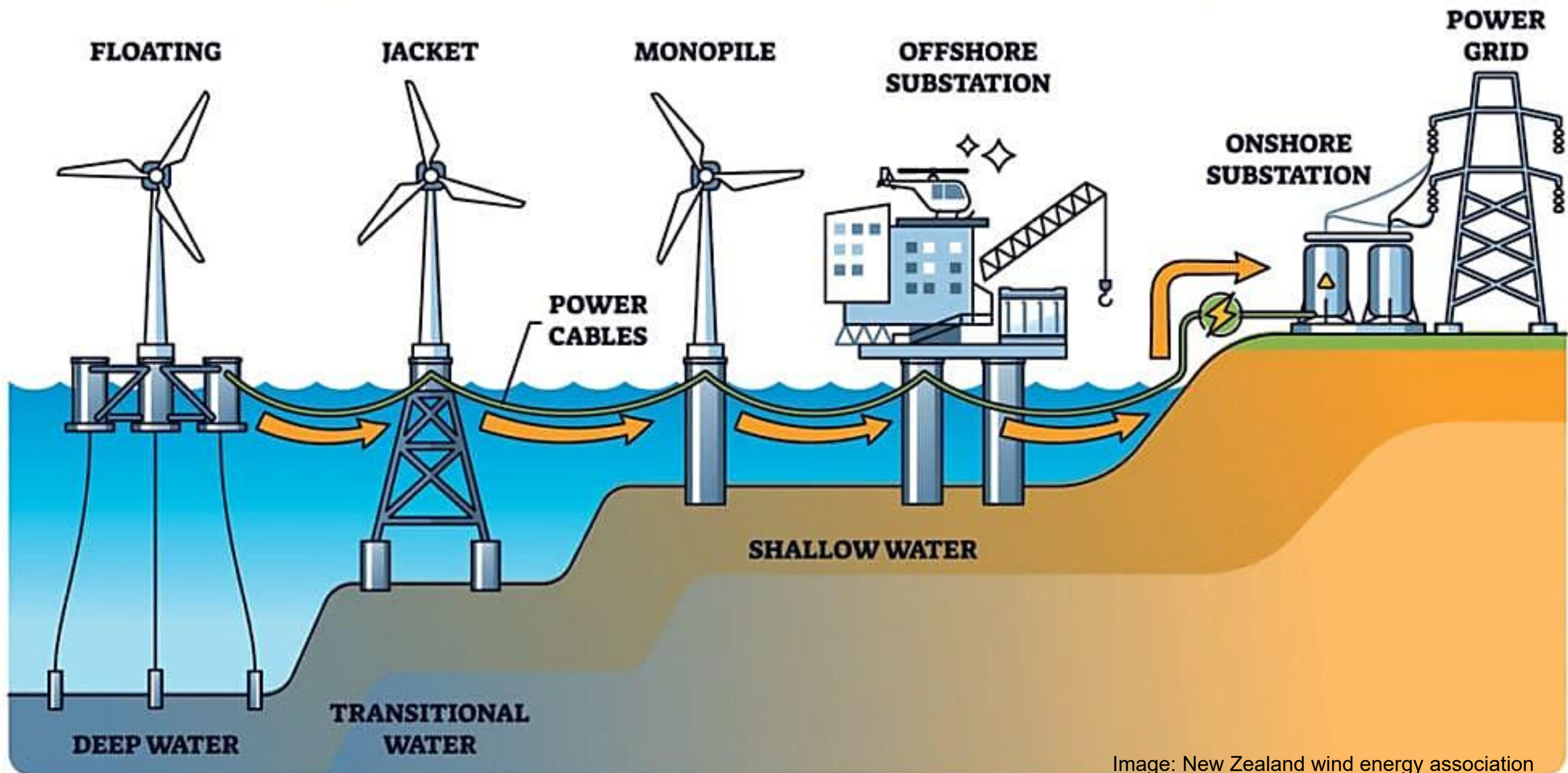


Image: New Zealand wind energy association



The Industry Shift

Maturing markets and disappearing subsidies demand new optimization technologies. The industry has shifted to data-driven models to optimize new turbine design and improved O&M strategies.

From assumptions to reality

- Traditional design models rely on conservative industry codes and generalized assumptions to account for uncertainties

Material fatigue resistance

Fatigue damage modeling

Long-term environmental conditions

Aerodynamic and hydrodynamic loads

Global dynamic response

Long-term operational conditions

Designed vs as-built tolerances

Unforeseen events

...

From assumptions to reality

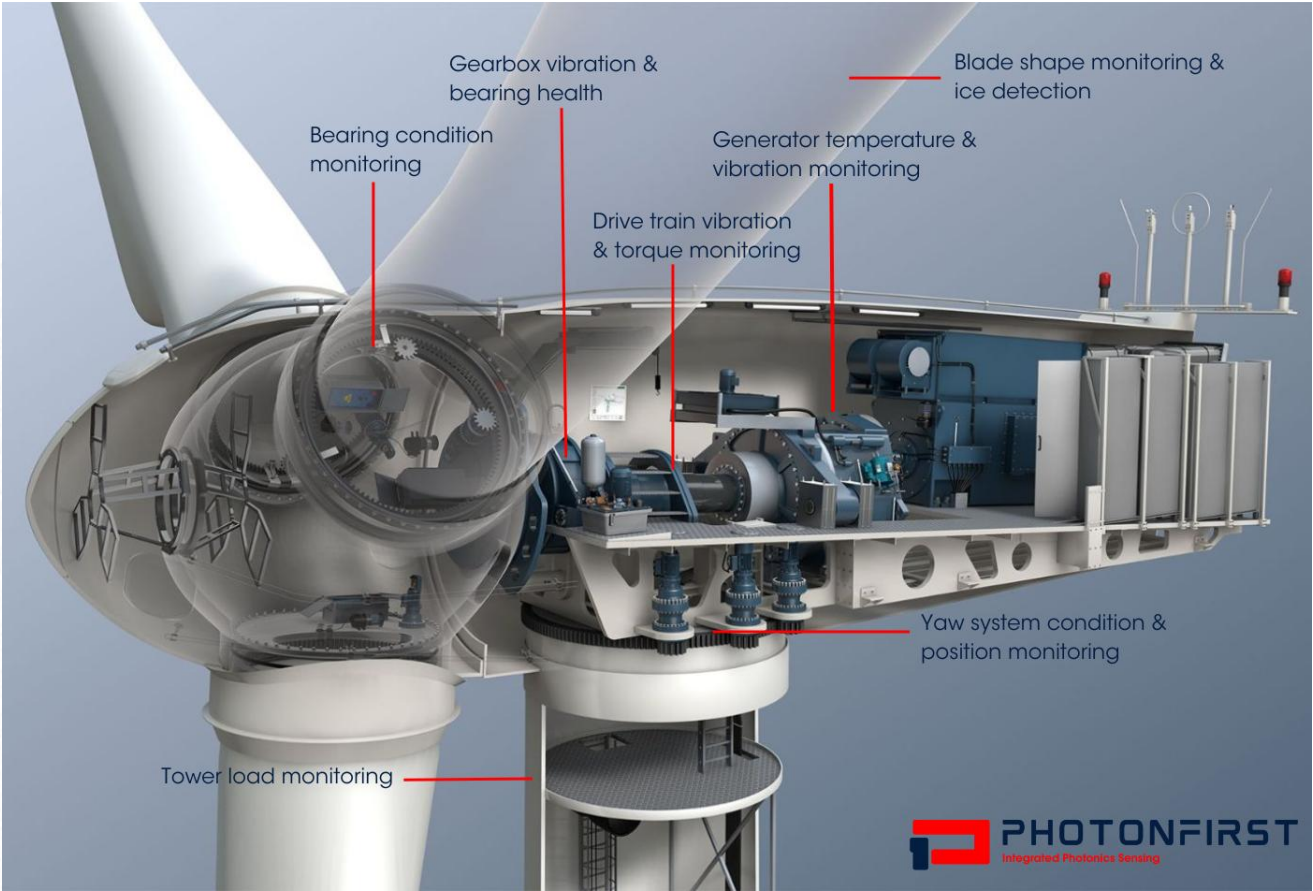
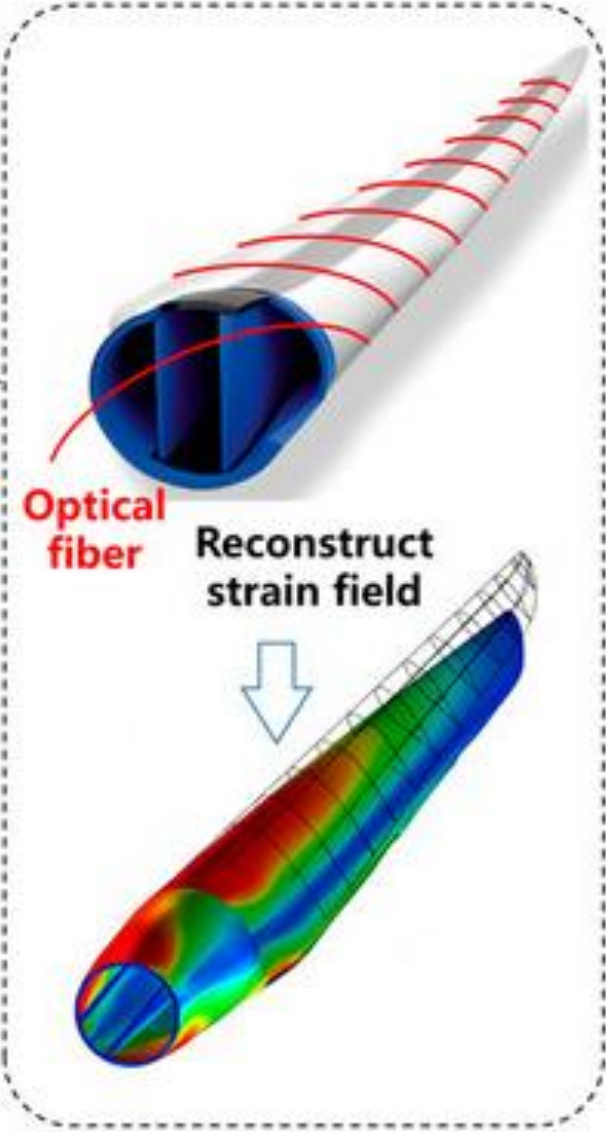
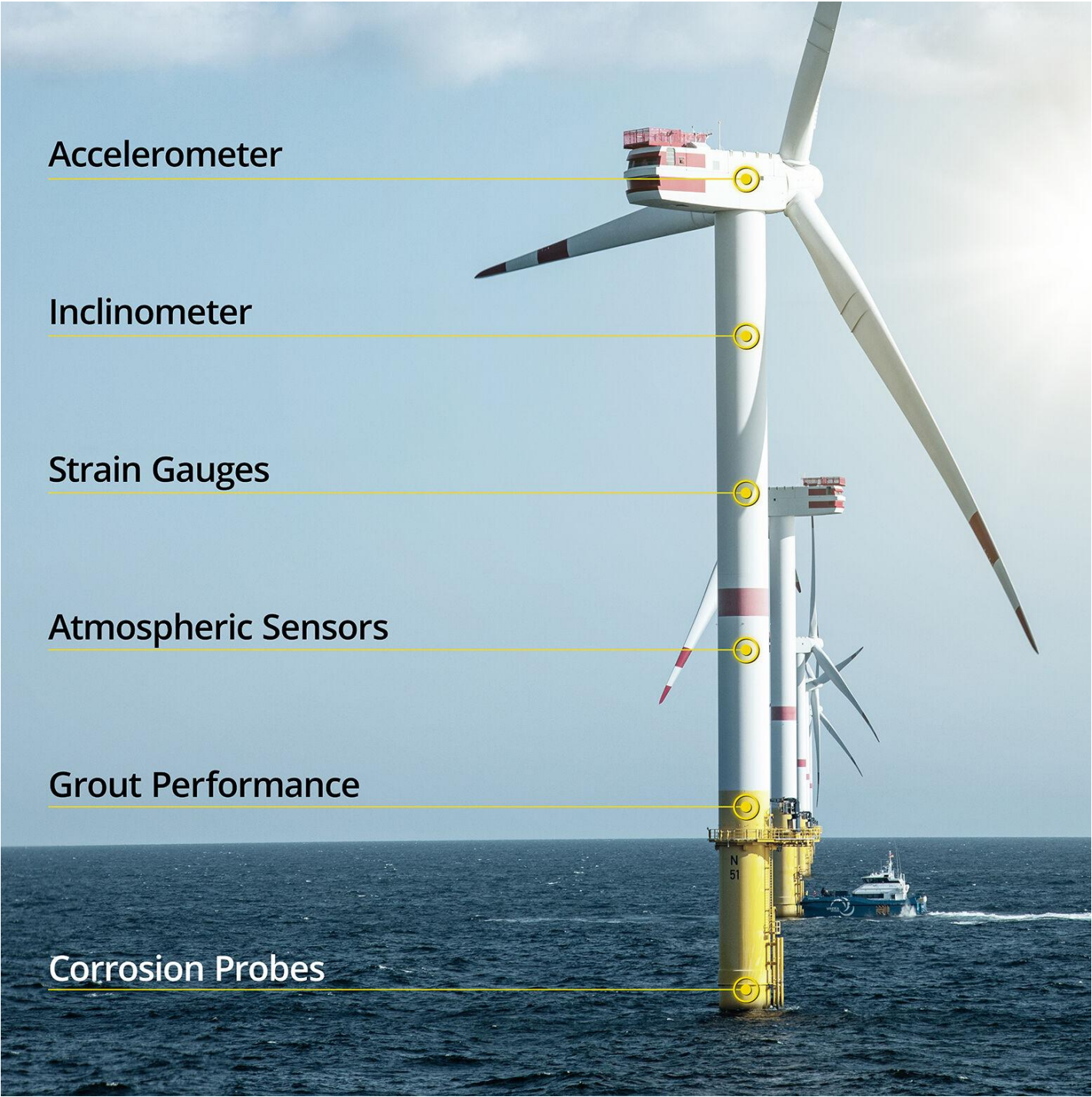
- Traditional design models rely on conservative industry codes and generalized assumptions
- State-of-the-art digital twins can replace estimates with real-time measurements from physical sensor
- Operational and environmental data are merged to form comprehensive monitoring solutions
- The system captures the operational behaviour of the entire fleet, down to individual turbines.



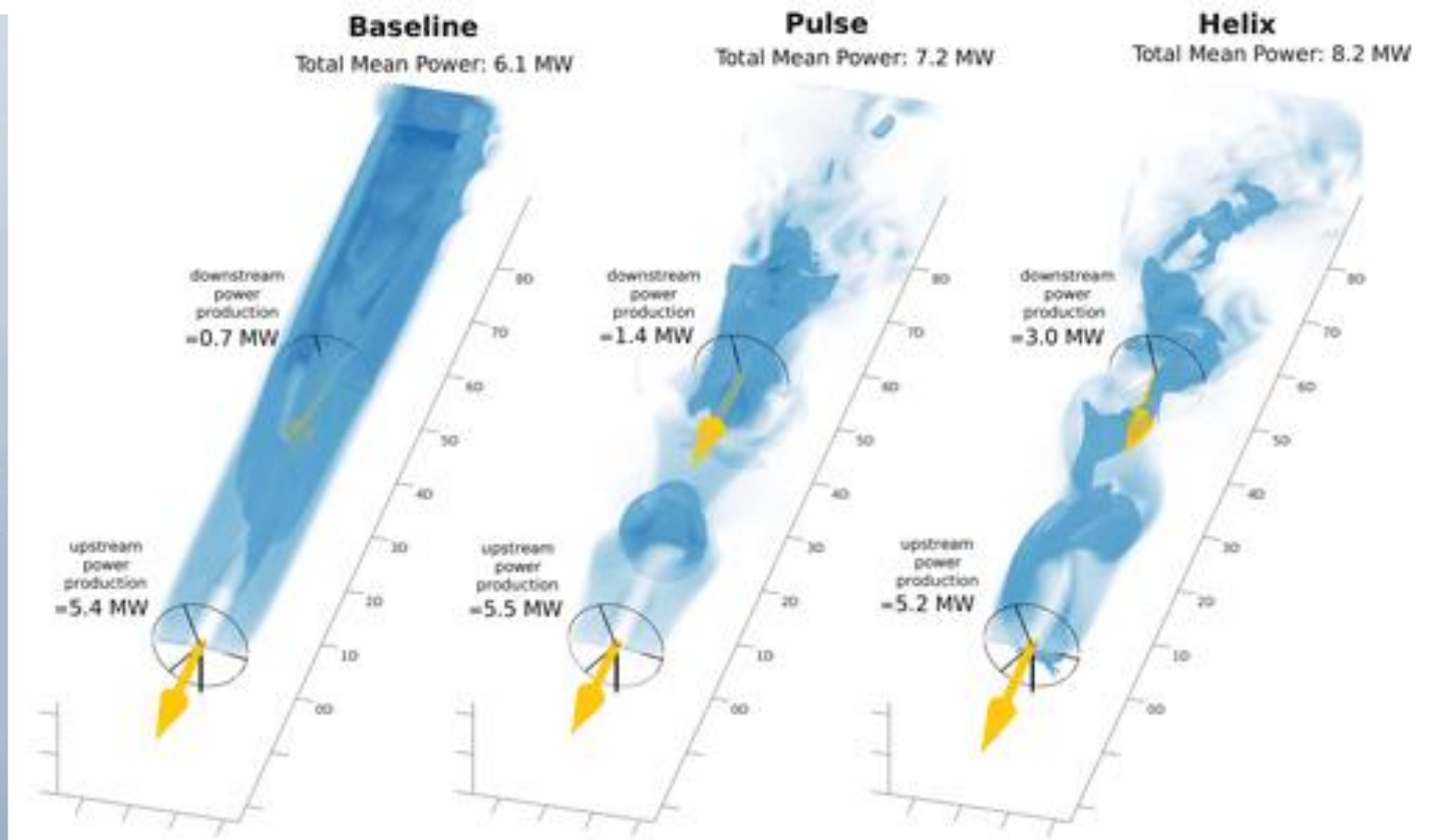
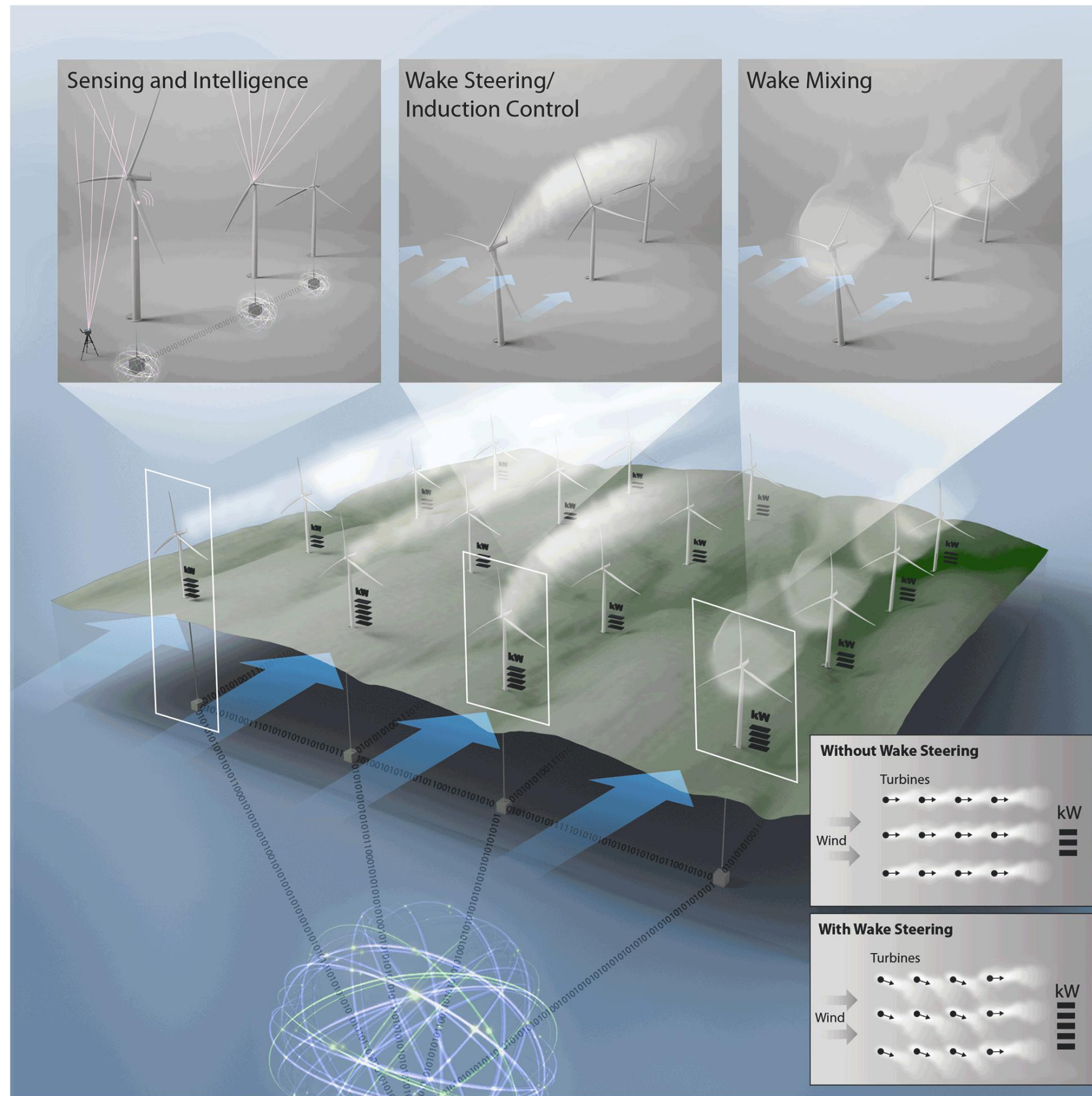
Capturing data for data-driven models of offshore turbines



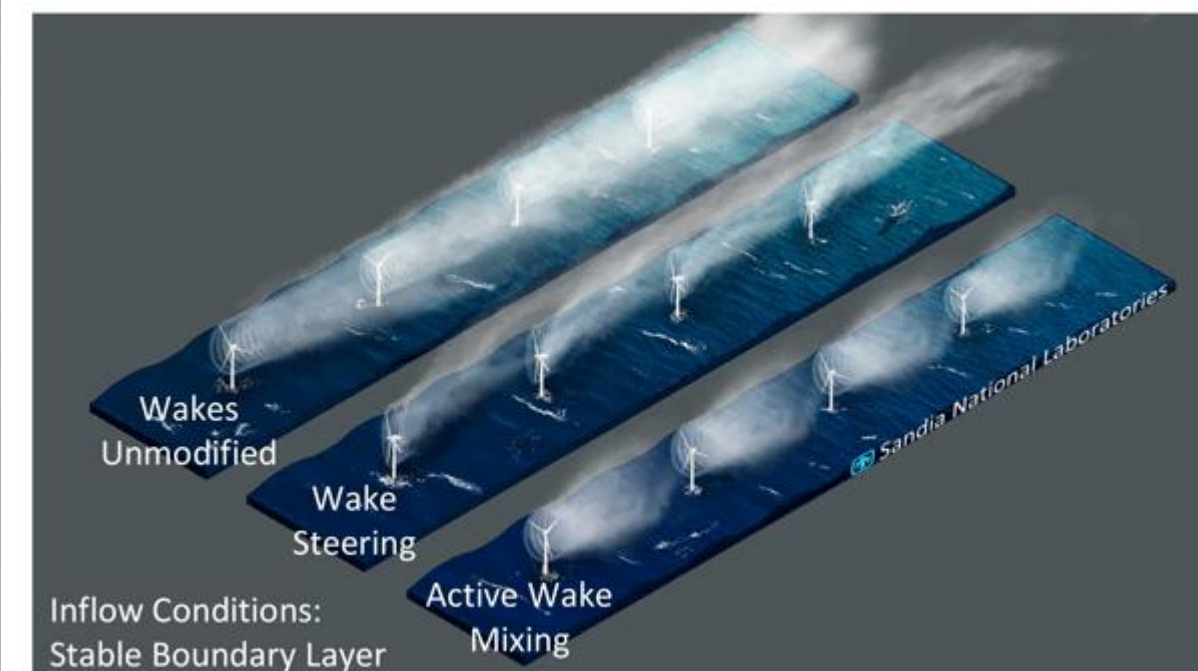
Structural health and load monitoring



Farm level sensing and control

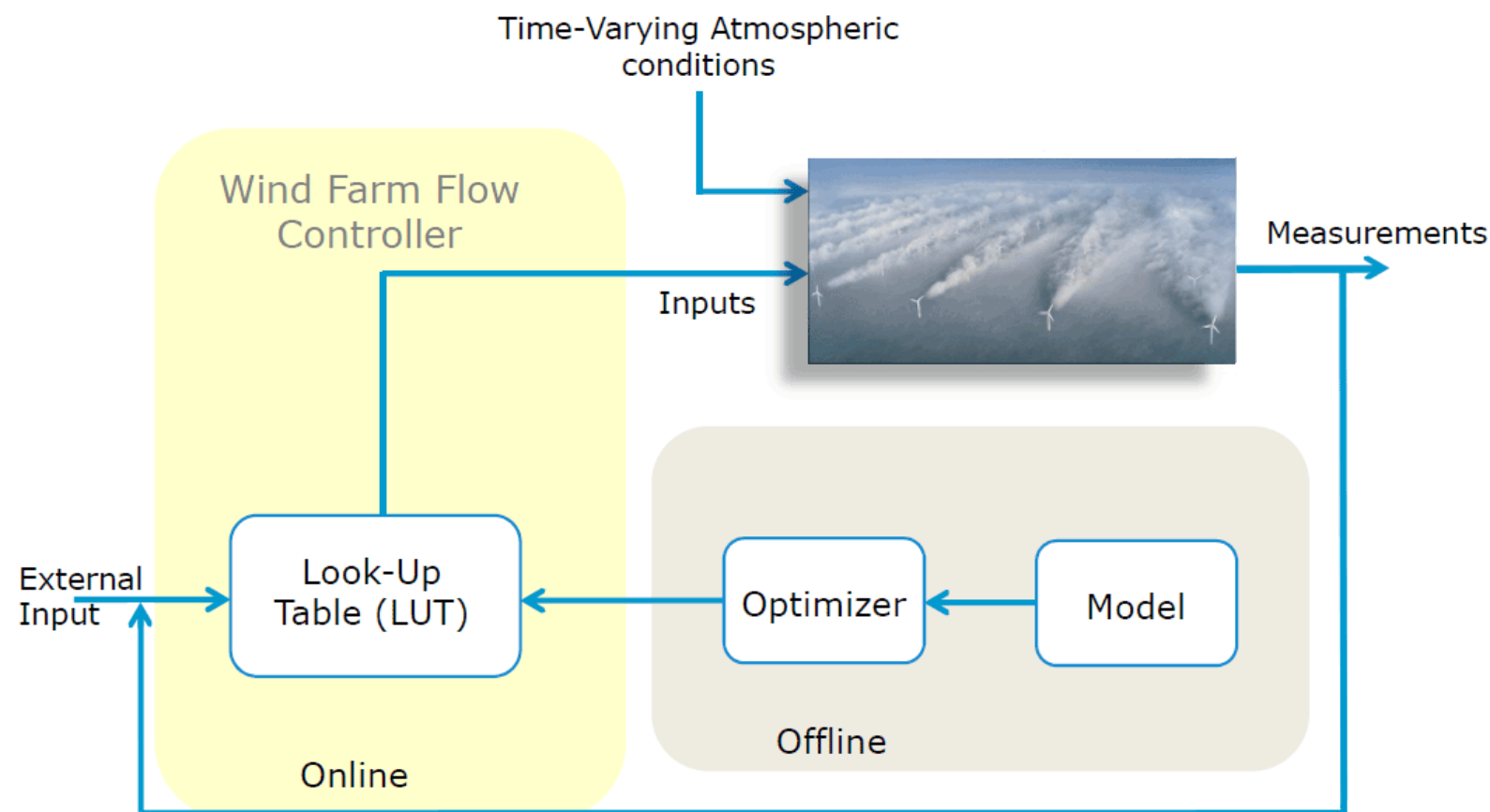


Meyers, Johan, et al. "Wind farm flow control: prospects and challenges." Wind Energy Science Discussions 2022 (2022): 1-56.
<https://doi.org/10.5194/wes-7-2271-2022>



Wind farm flow control

Current practice open-loop control

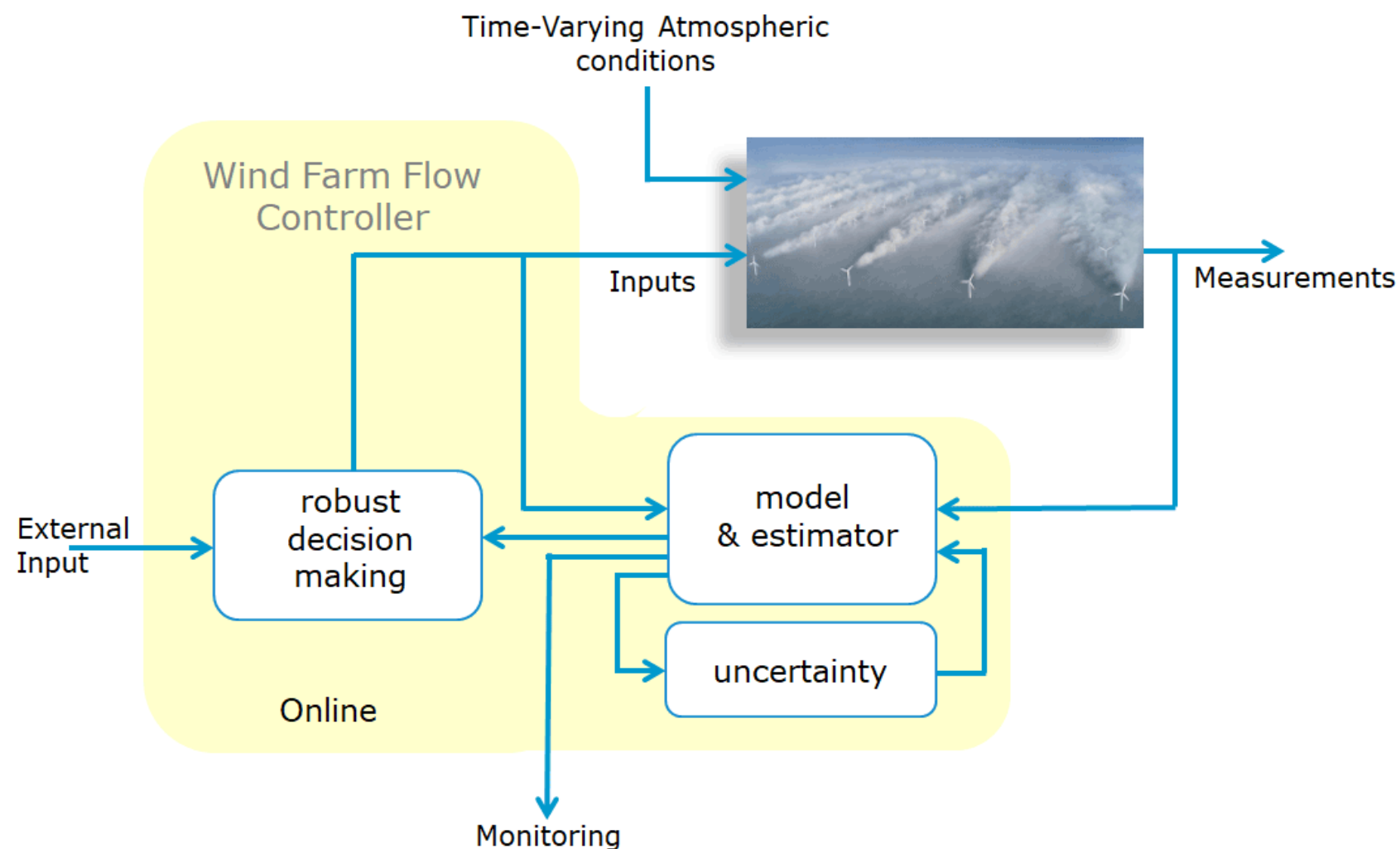


- Offline-calibrated, steady-state models
- Optimized for different wind directions, wind speeds, turbulence conditions, etc. which are fed into a LUT with optimal control settings
- If the model misunderstands the wake interaction between two specific turbines, the wind farm will consistently lose power efficiency under those specific wind conditions.

Performance strongly depends on the model accuracy and is prone to unmodeled disturbances

Wind farm flow control

Future wind farm flow control Closed loop-control



- Real-time optimization framework that accounts for model uncertainty and unknown uncertainties
- Allows active wake steering, minimizing impact on neighboring turbines

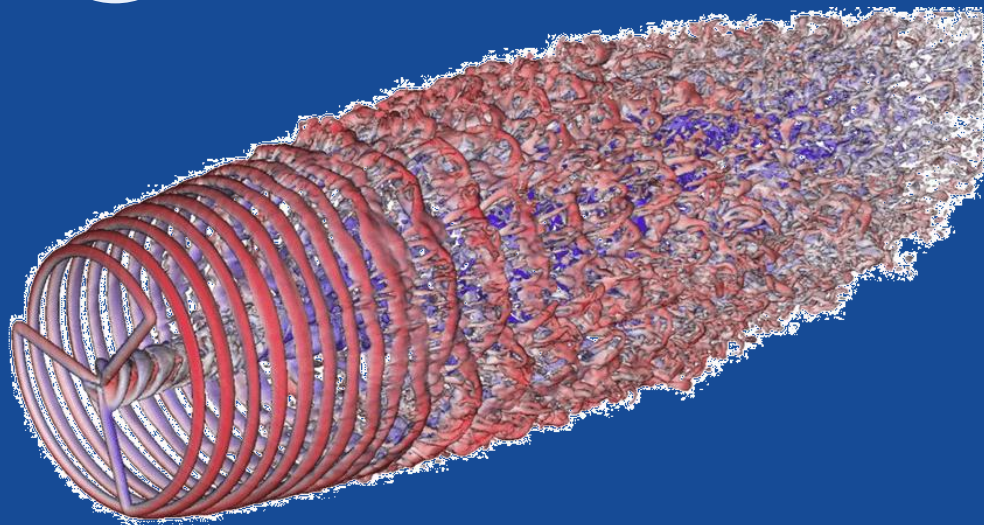
Result: data-driven calibrated models that can be used for real-time decision making = digital twin

Target: Increasing AEP → how does it affect the individual assets?

From aerodynamic wakes to structural loads

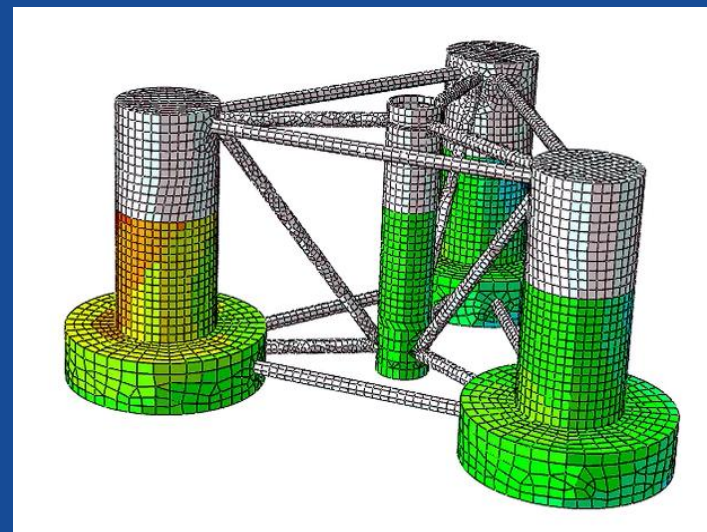
Tracing the causal chain from farm-level flow control optimization to component-level fatigue life consumption.

1 Farm level



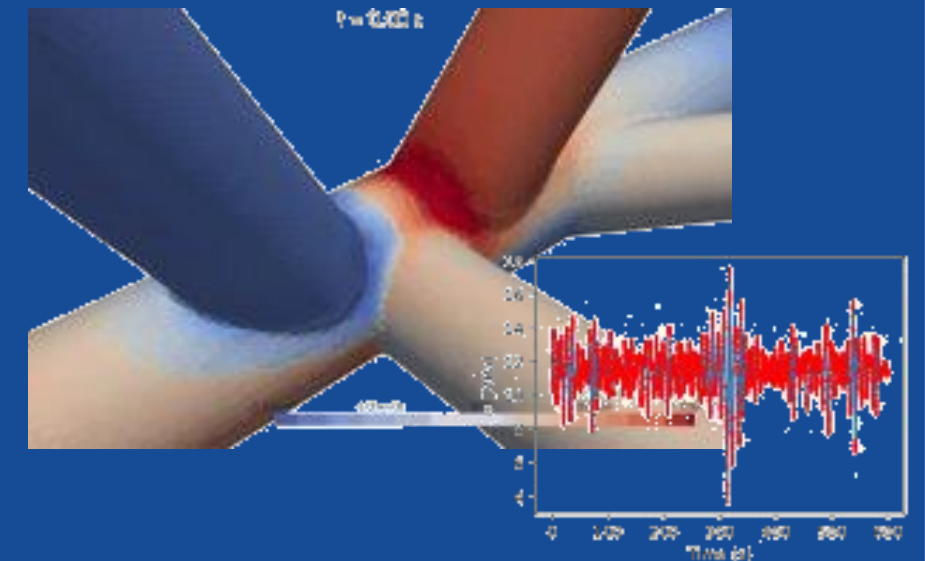
Wake steering optimizes farm-level yield but fundamentally alters inflow conditions, creating asymmetric rotor loading and wake-induced turbulence for downstream assets.

2 Turbine level



Aerodynamic variations translate directly into dynamic load amplification, propagating cyclical bending moments and thrust variations throughout the entire support structure.

3 Component level



Dynamic loads dictate the fatigue governing stresses at weld imperfections, leading to fatigue degradation at critical structural joints.

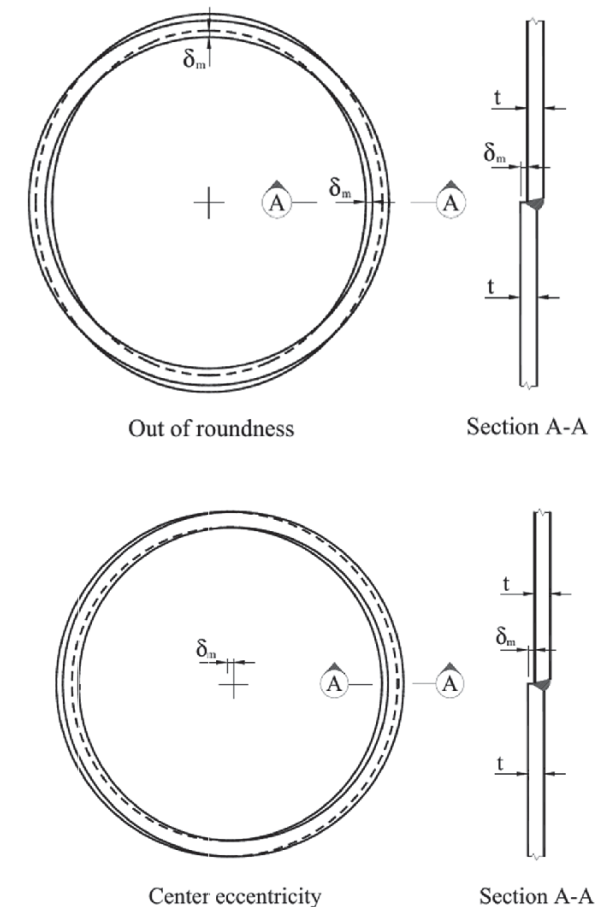
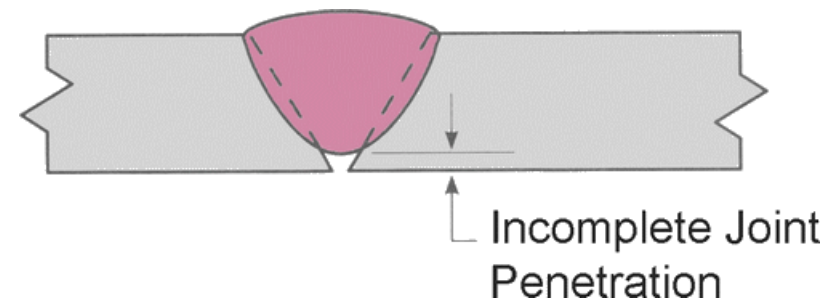
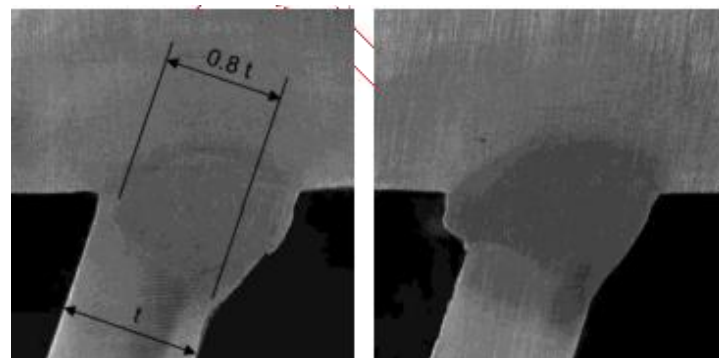
Rethinking fatigue resistance models: The challenges associated with *perfect* models



The challenge of fatigue in critical structures

As-design models are idealized, but reality is not. How do we bridge this gap?

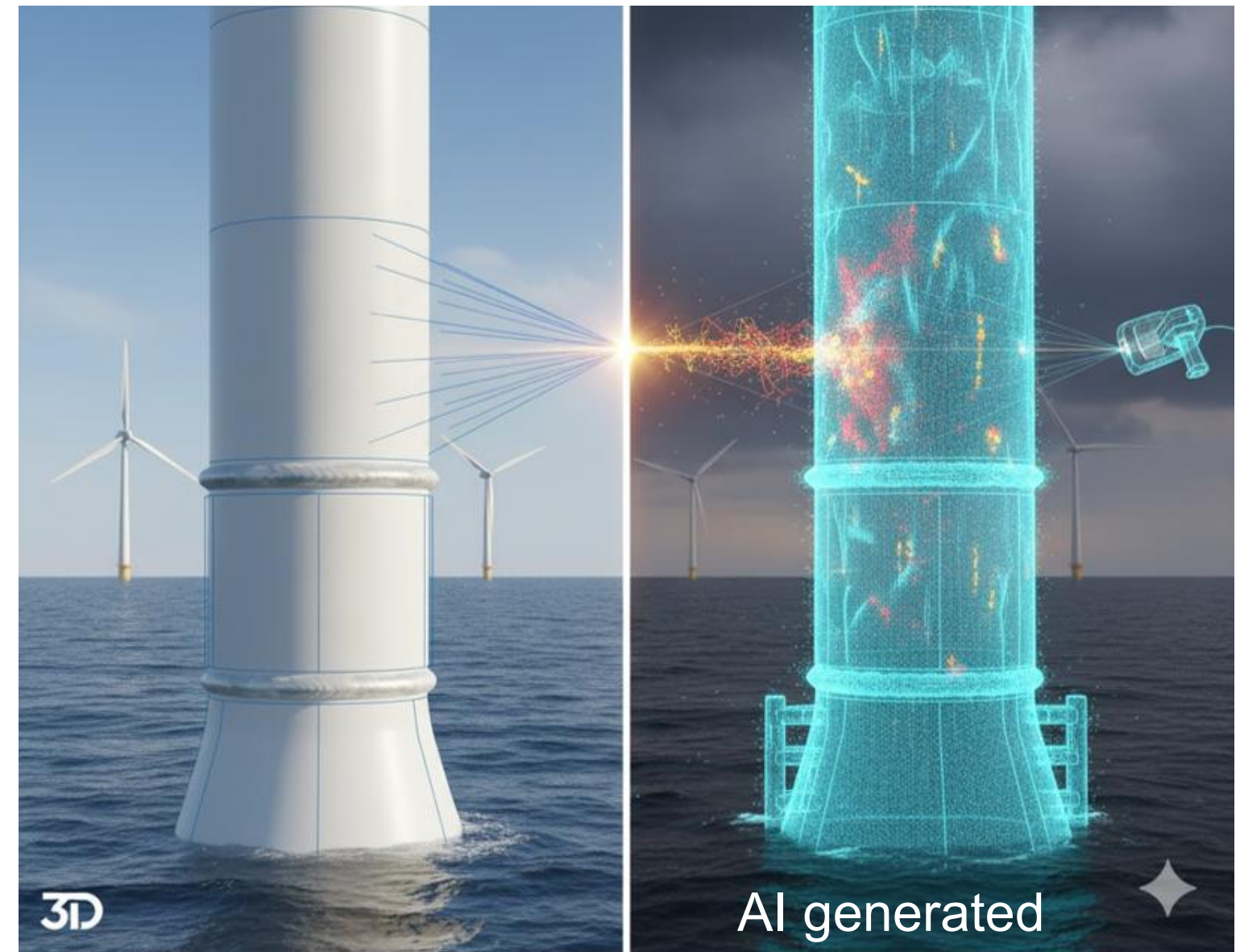
As-built structures differ from the as-designed (ovality, misalignment, weld profile, incomplete weld);



The challenge of fatigue in critical structures

As-design models are idealized, but reality is not. How do we bridge this gap?

A new approach: from
idealisation to reality
We create **exact** digital
copies the structure



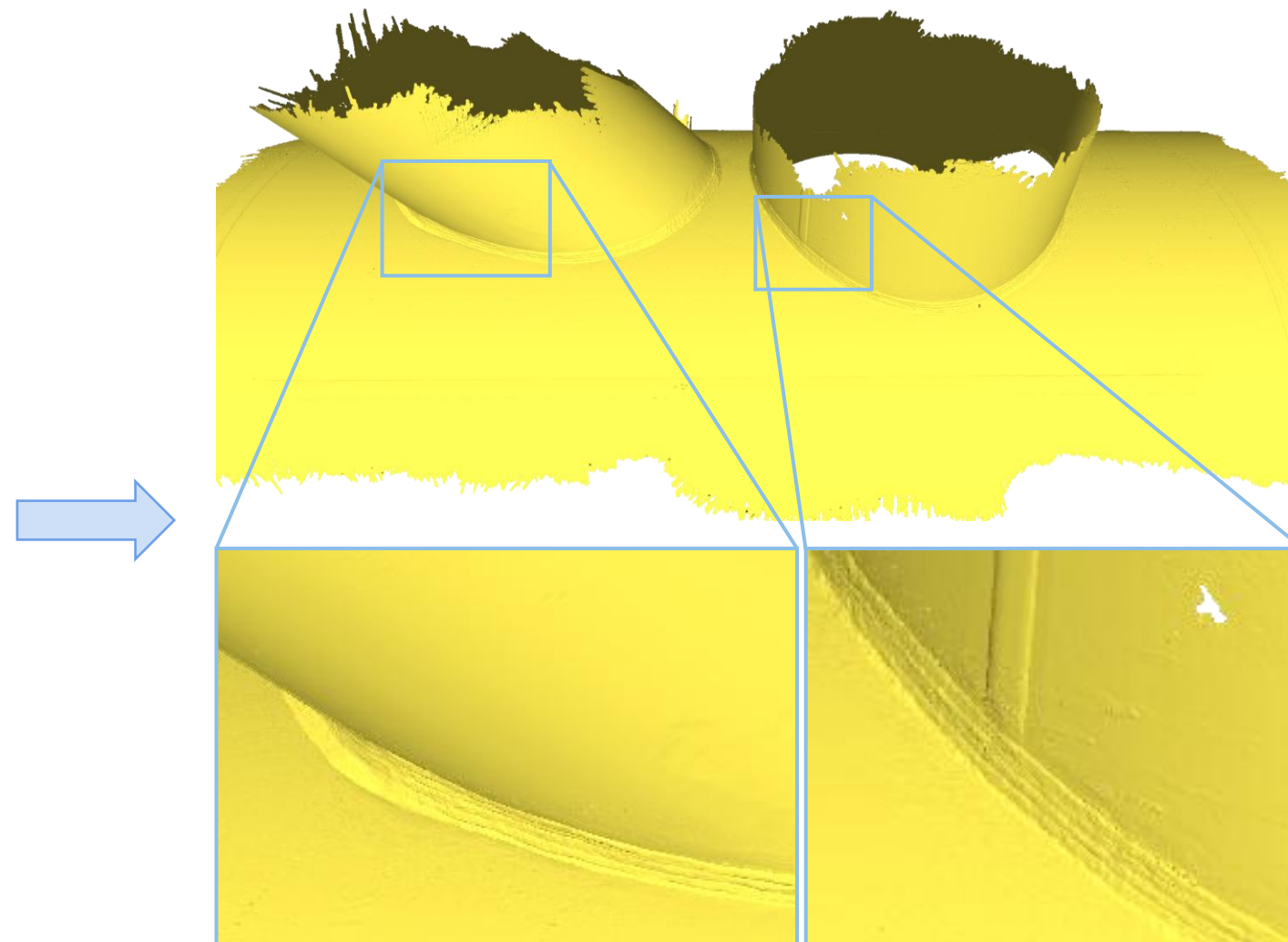
Capturing the 'as-built' reality with 3D scanning

Moving beyond idealised designs: We measure what is really there.

3D Laser Scanning: Capturing the overall macro-geometry of complex industrial joints.



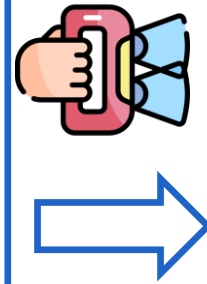
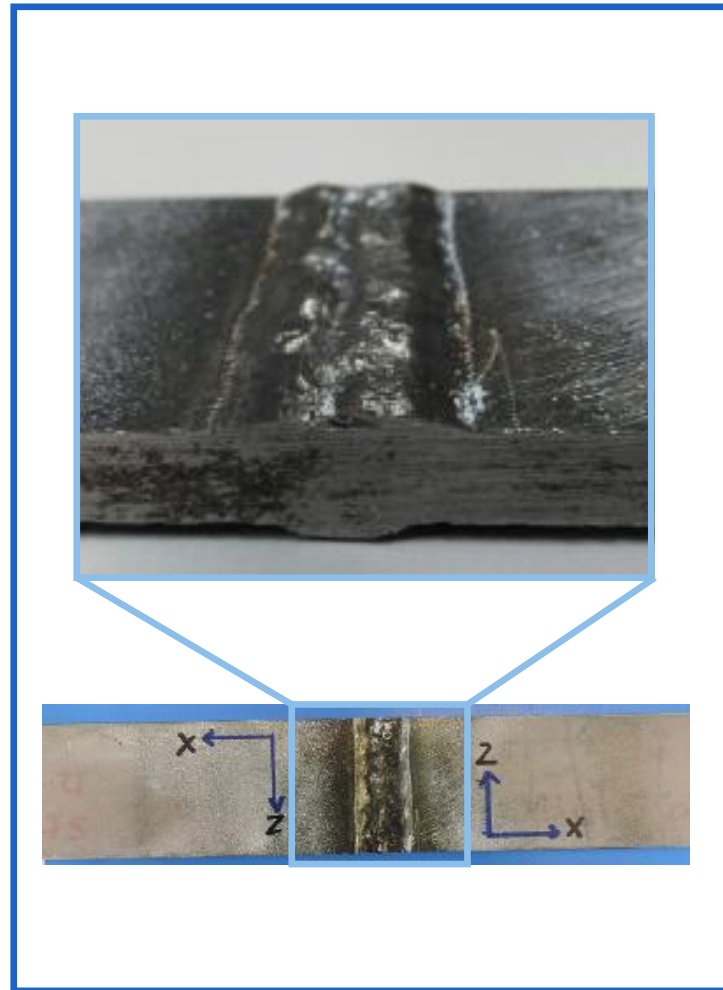
Point cloud: Raw output of the scanner.



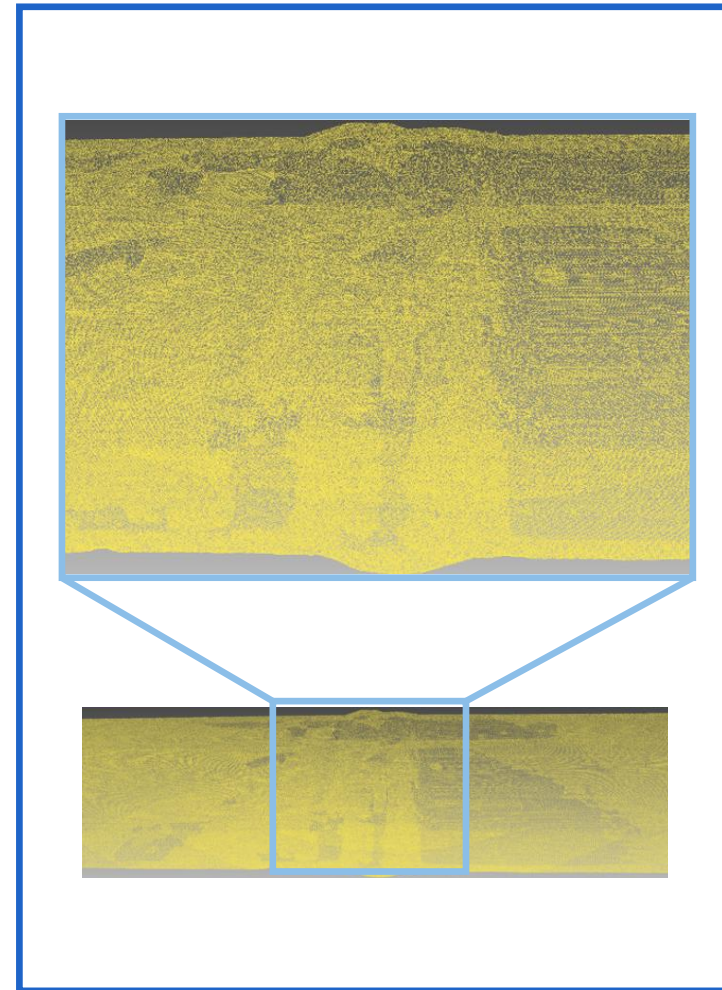
From point cloud to simulation model

Overall workflow of the developed framework

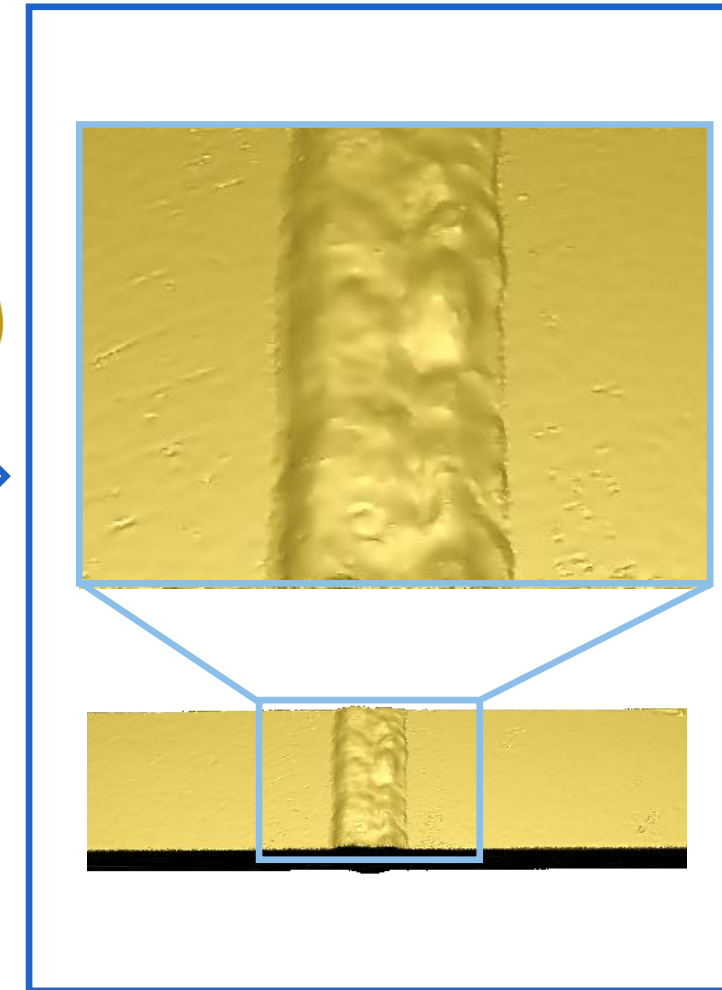
1. Welded joint



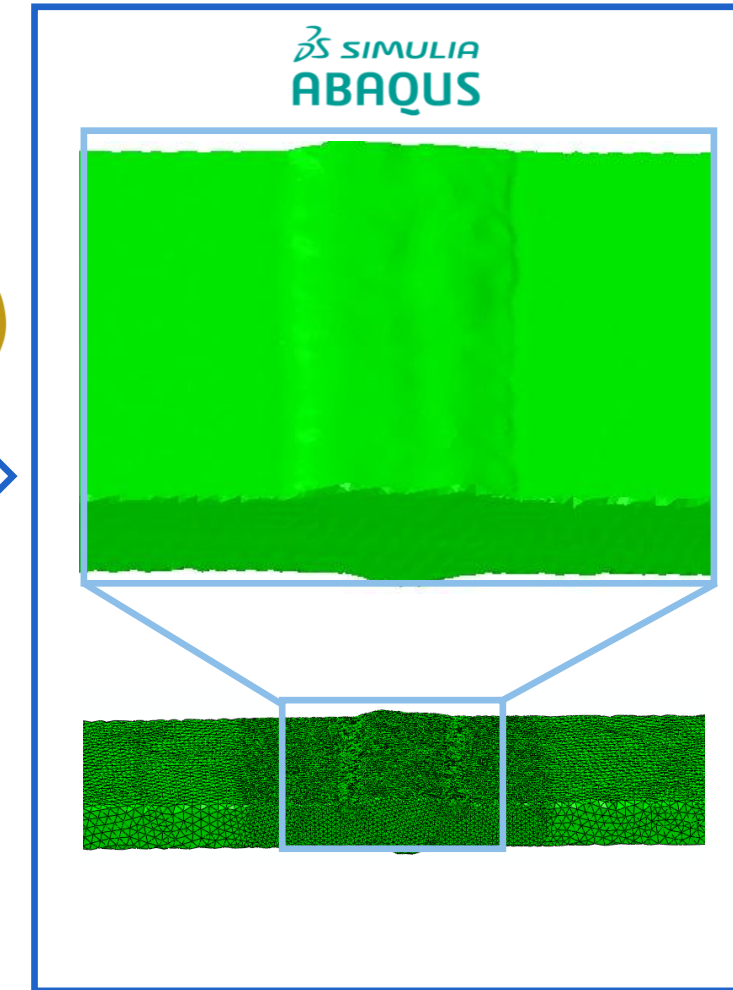
2. Point cloud data



3. Surface mesh (.stl file)

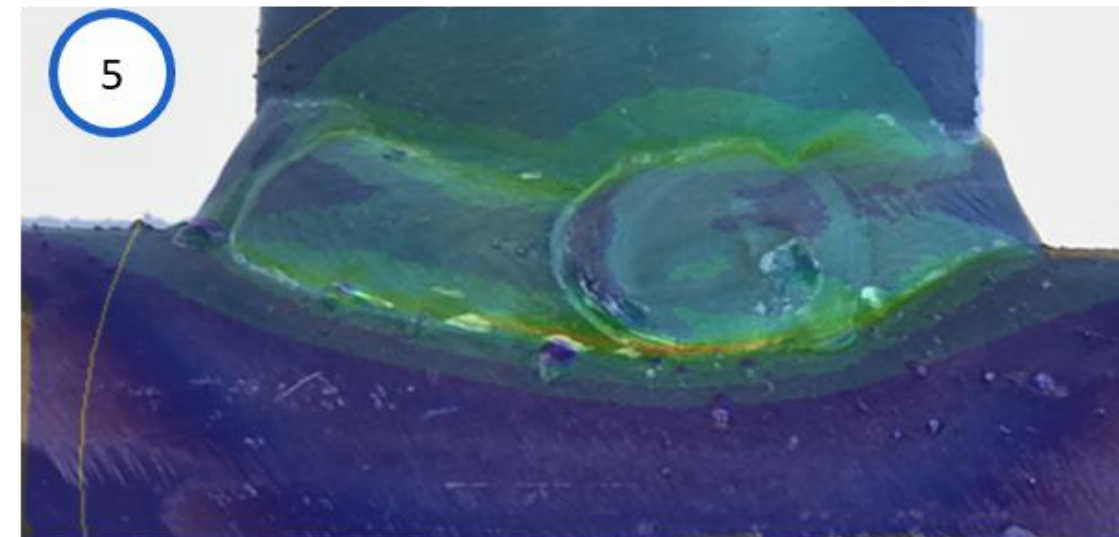
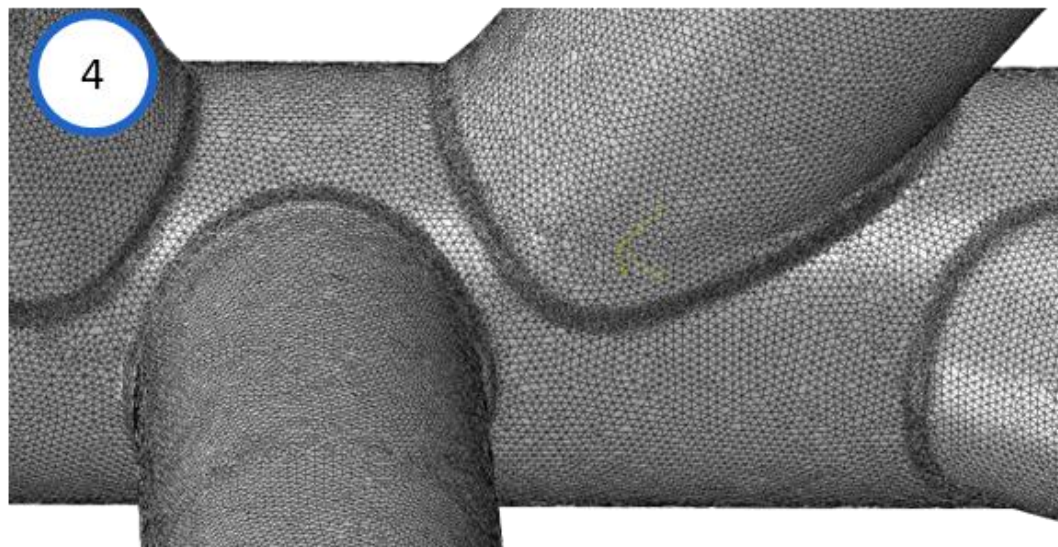
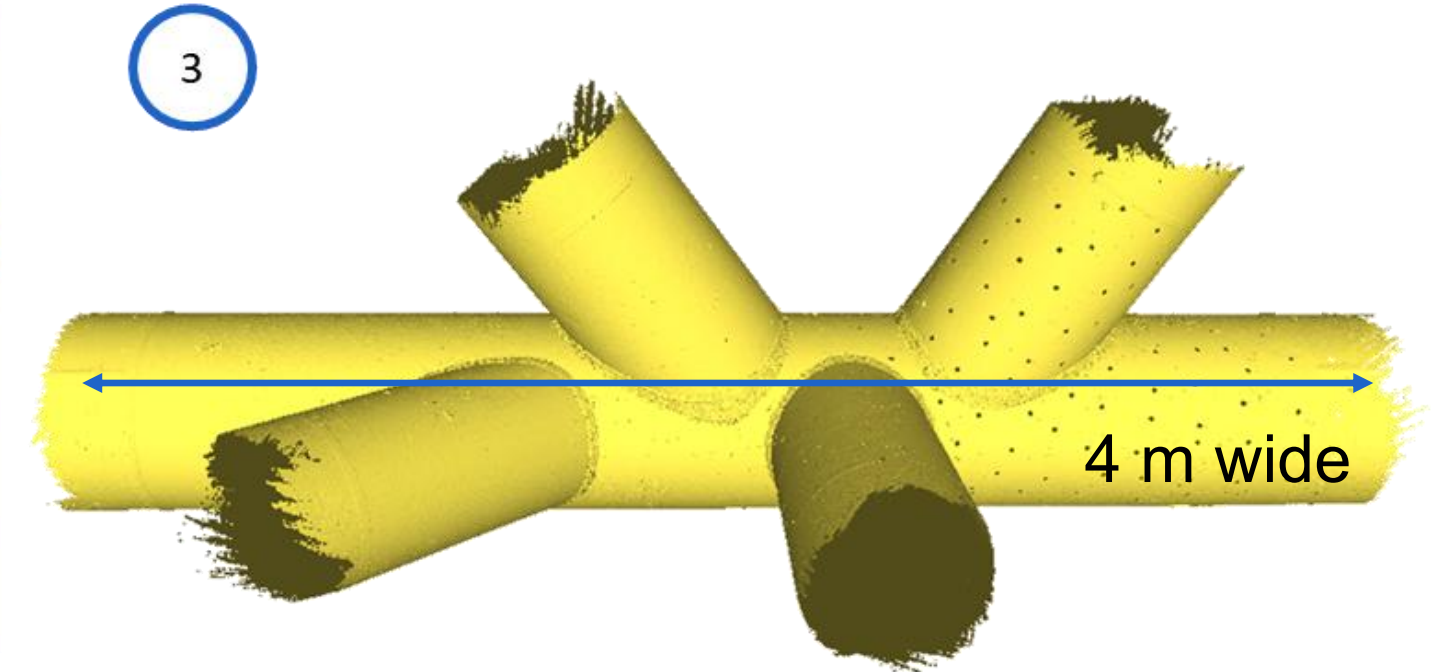


FE model



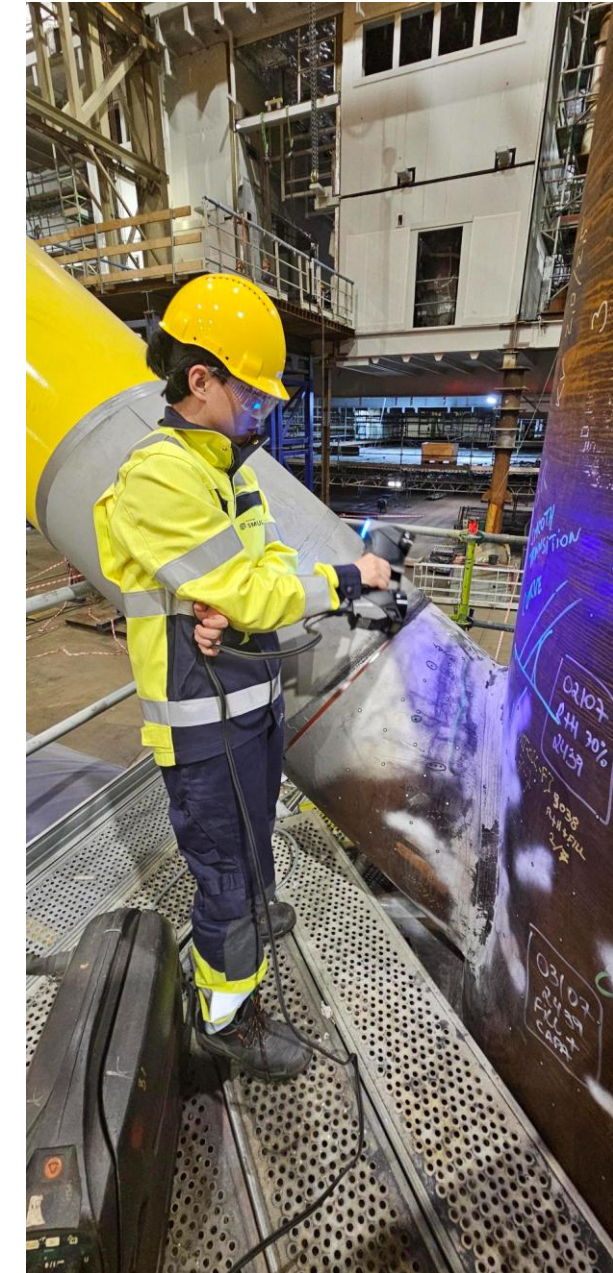
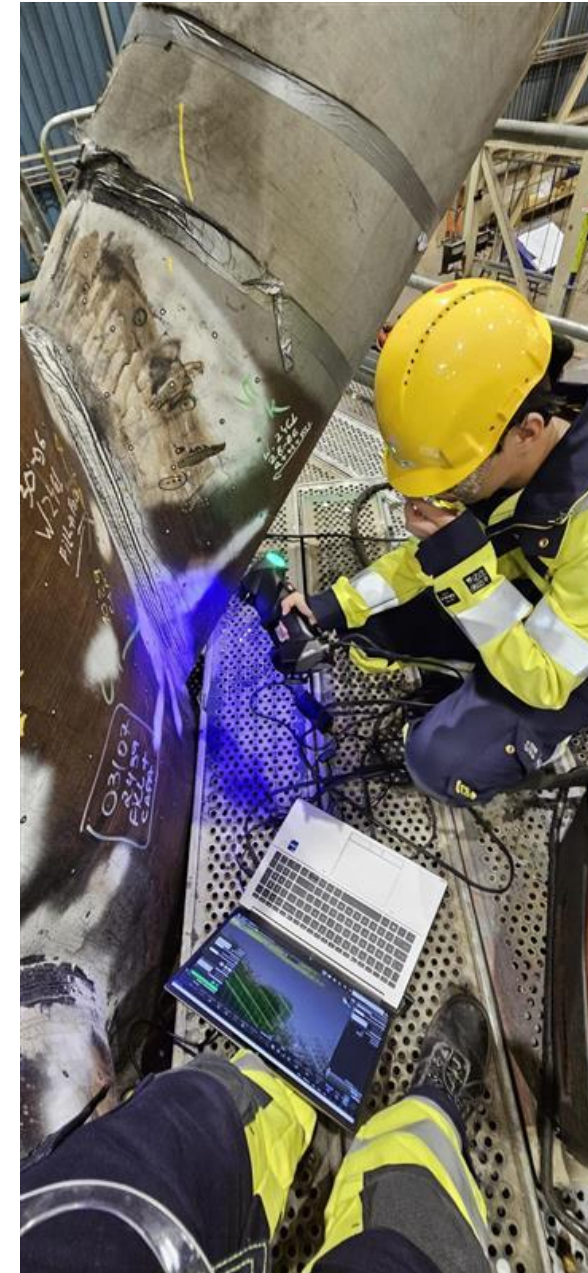
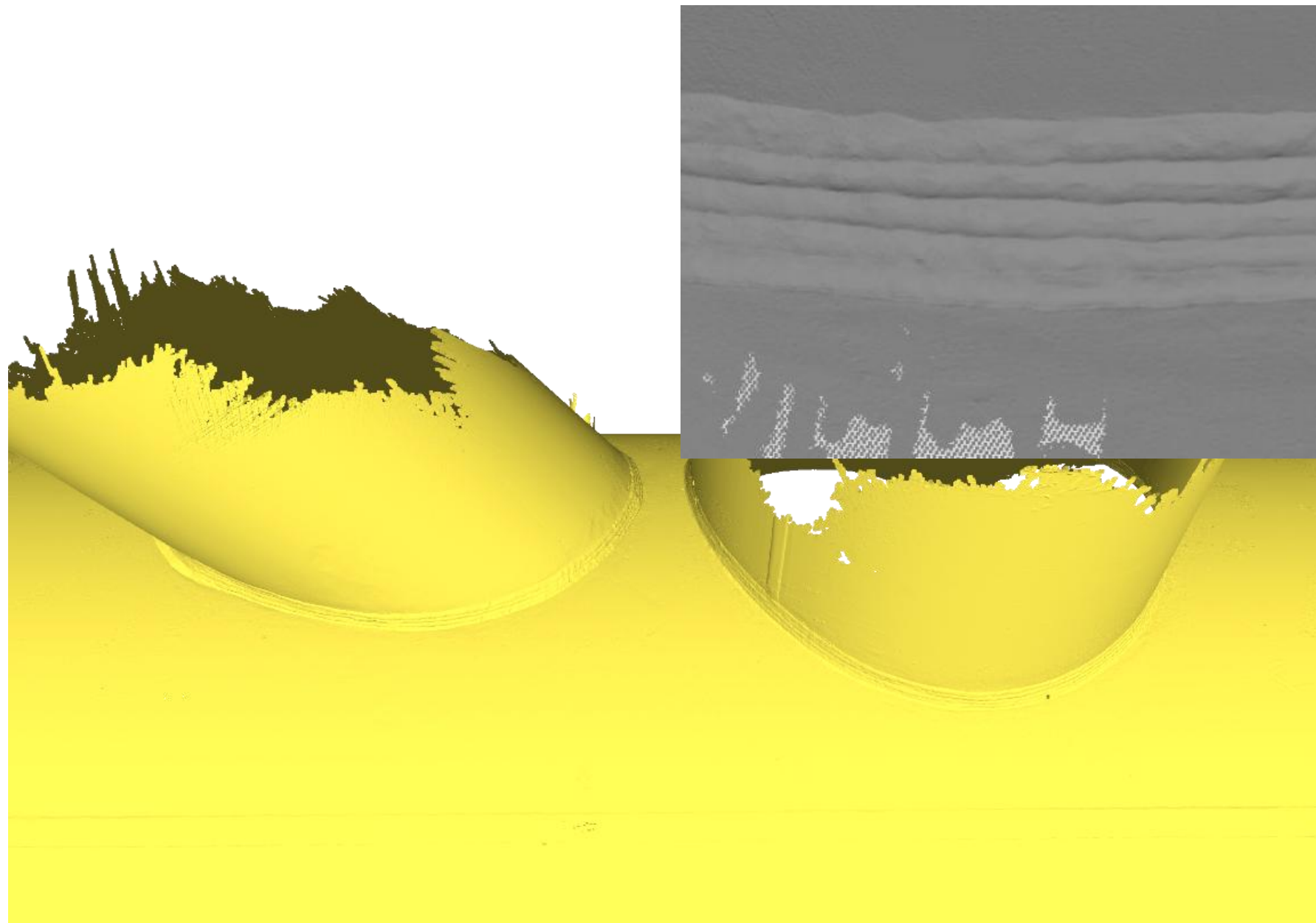
Effect of as-built geometry on fatigue strength

Application to a full-scale offshore joint



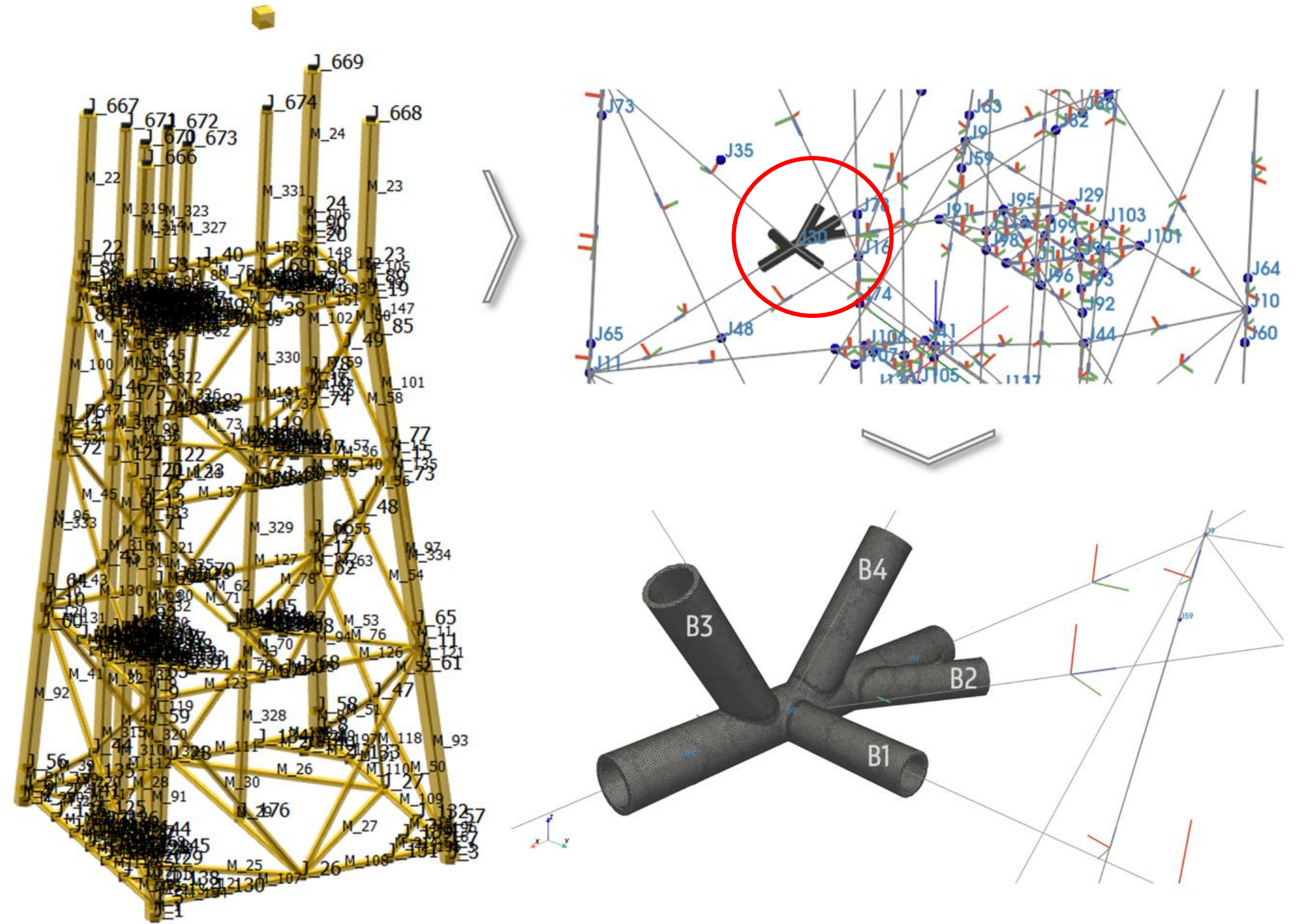
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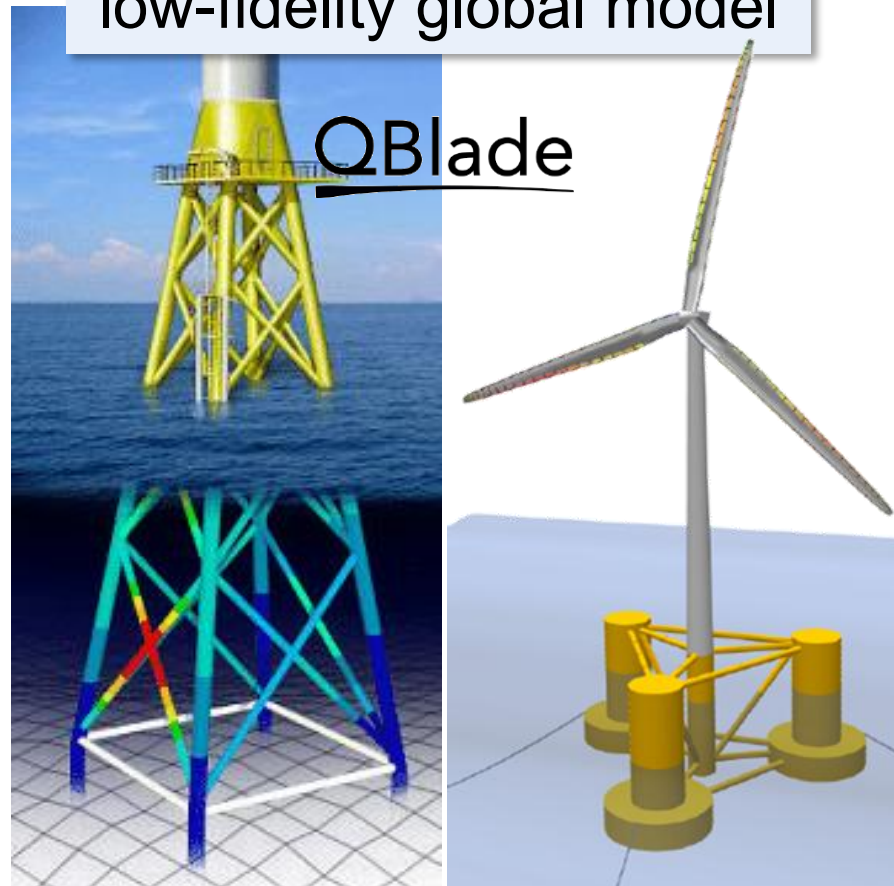
Coupling high-fidelity model to global model

- Global low-fidelity model for calculation of overall structural response
- Global-to-local coupling introduces structural loads onto the local finite element model



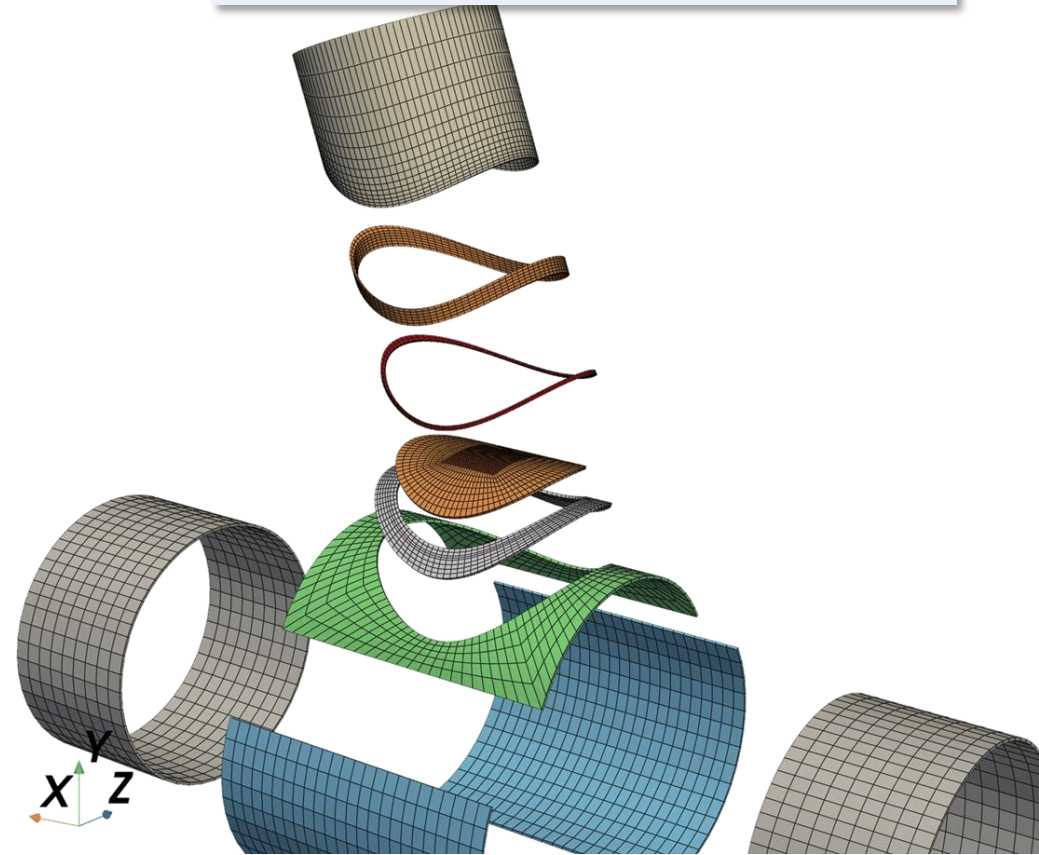
HFJoint: High-fidelity wind turbine simulations

Aero-hydro-servo-elastic
low-fidelity global model

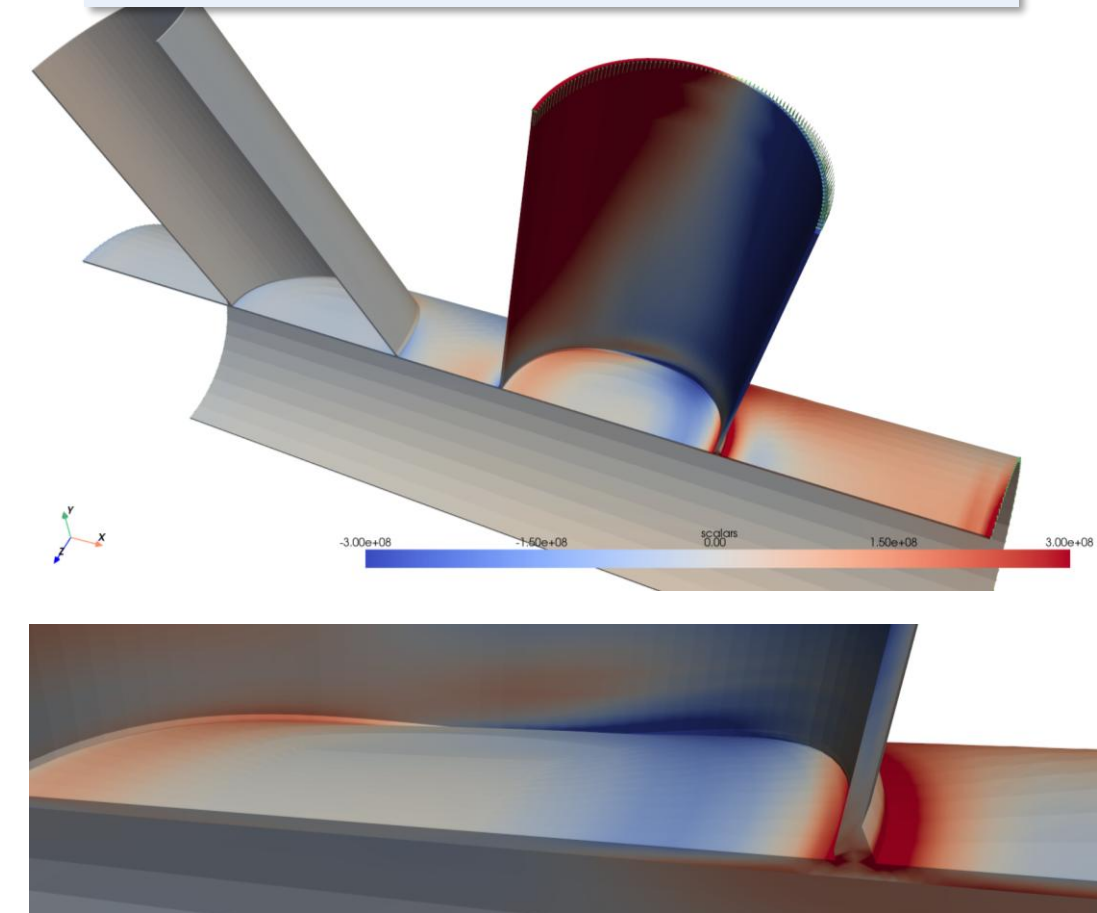


QBlade

Automatic load mapping
to auto-generated solid
element submodel



In-house developed lightweight
structural FEM solver



General framework for structural DT

General framework

Goal achieved: reduced scatter on fatigue life assessments through the use of as-built representation of the structural joints and load spectra based on real measurements

But.. Too slow and reactive rather than proactive



Bringing it together: Towards real-time, health-aware turbine and farm level control



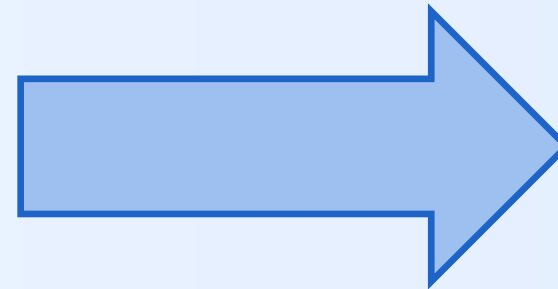
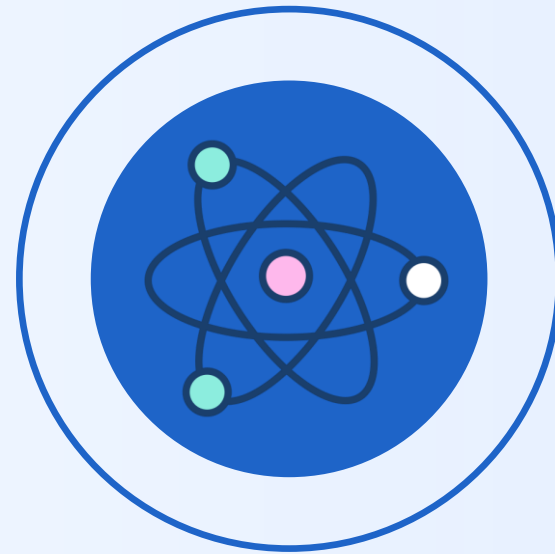
Digital twin maturity matrix: offshore wind

	Gen 1 Static / Design	Gen 2 Data-driven dashboards	Gen 3 Real-time hybrid
Data Flow	Unidirectional	Unidirectional	Bi-directional
Core technology	FEA / CFD / MBD	IoT sensors	PINNs + GNN/ROMs
Output	Post-mortem diagnostics	Reactive alerts	Real-time optimization
Extrapolation	None	Poor	Robust under varying conditions

The modelling paradigm

Pure physics-based

High fidelity, interpretable but computationally prohibitive



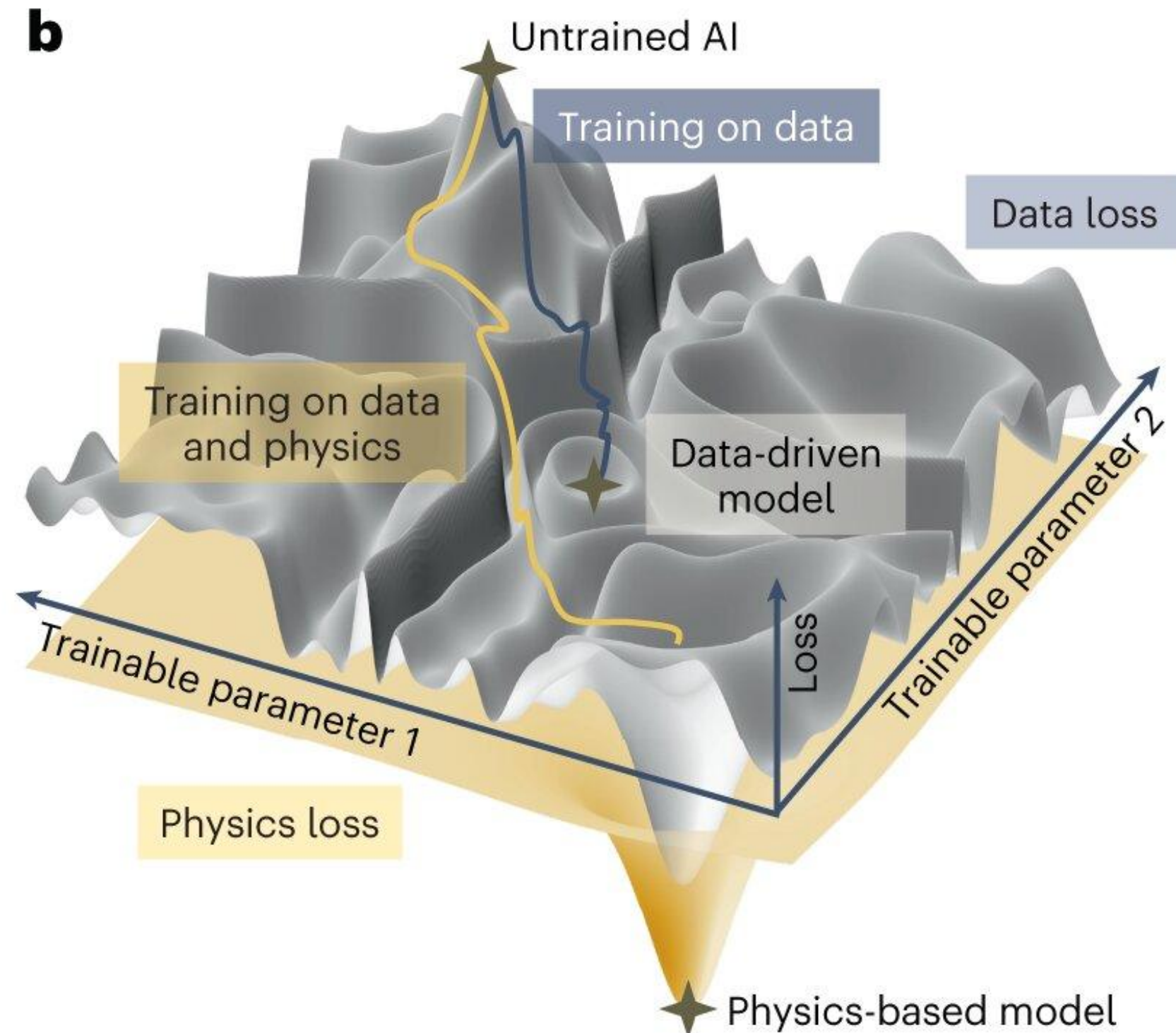
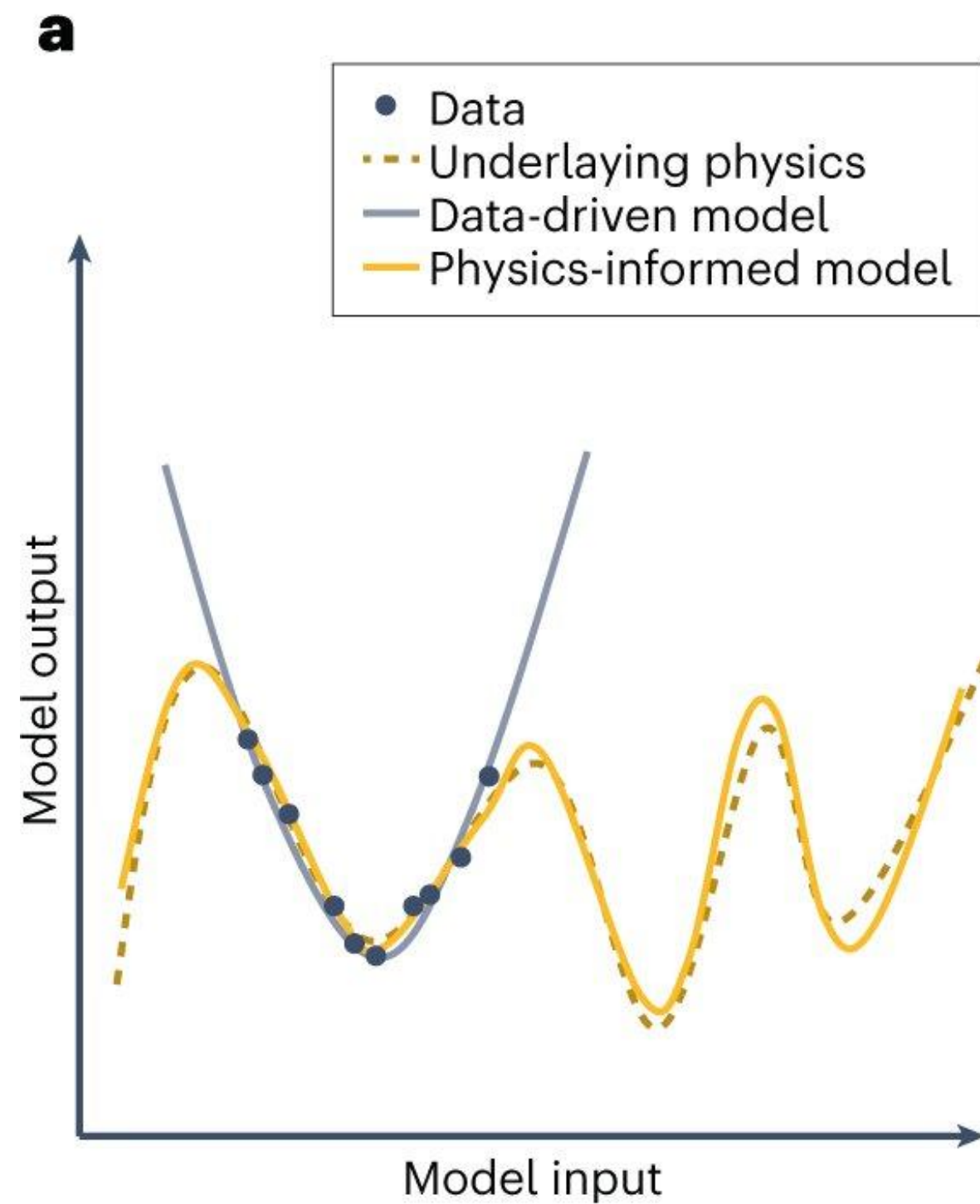
Hybrid AI-Physics
Fast, robust Interpretable

Pure data-driven

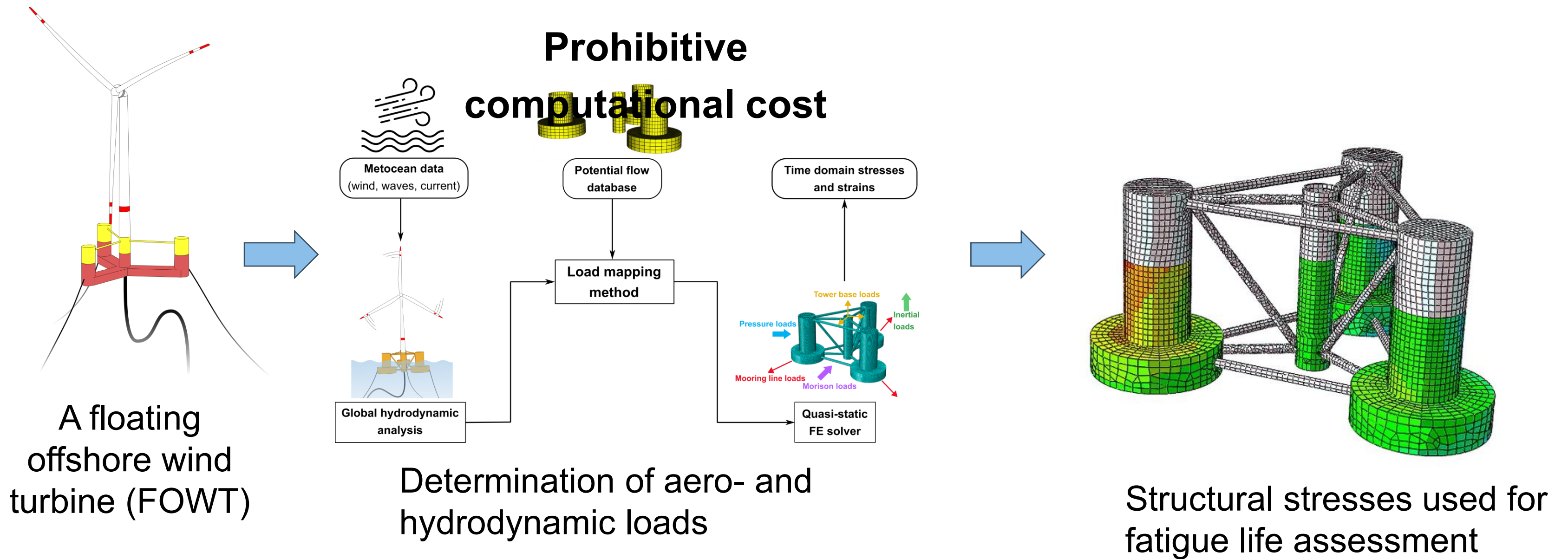
Fast, scalable, but black-box and physically inconsistent.



Physics-informed data-driven modeling

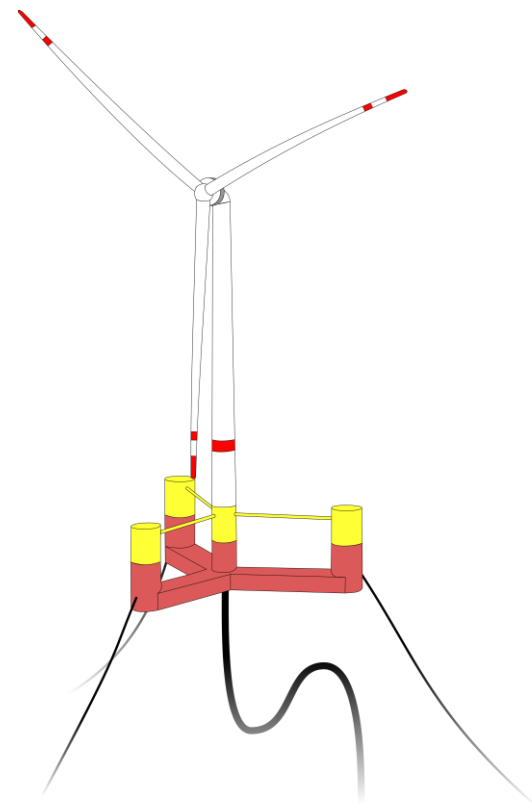


The computational barrier: determining real-time structural behaviour

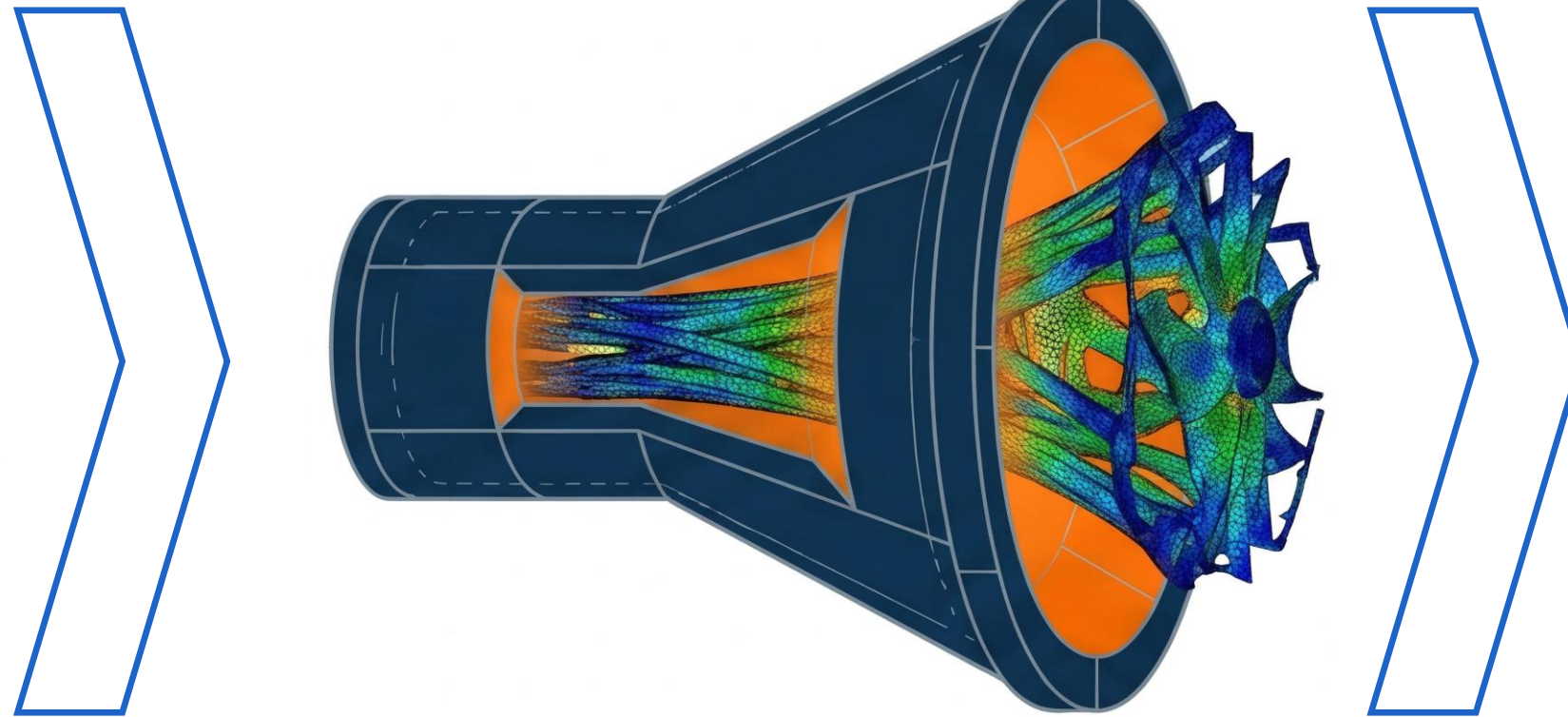


Breaking the computational barrier

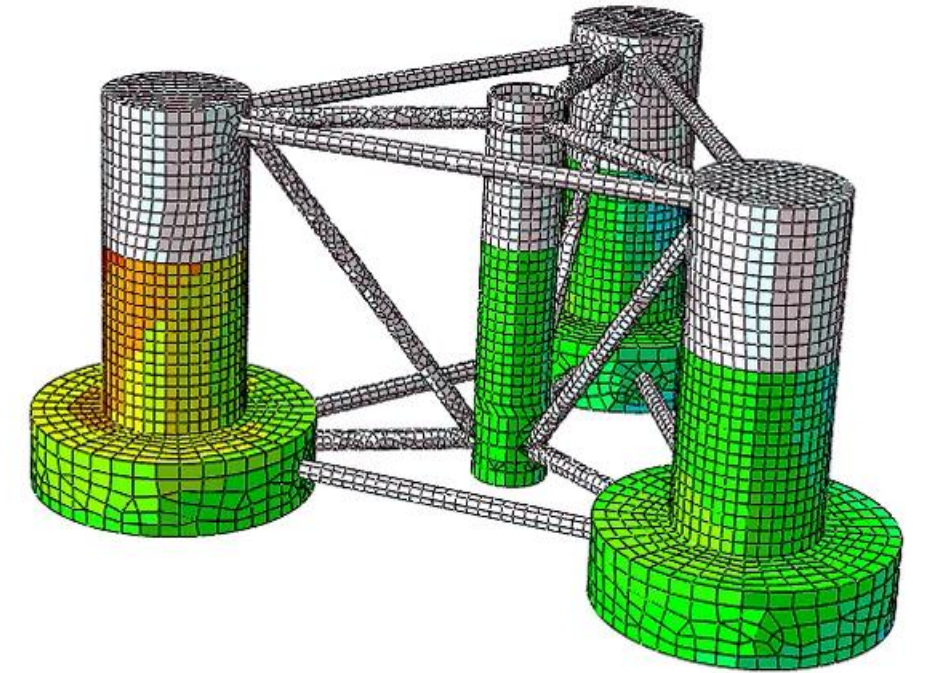
Reduced order models based on physics-informed neural-networks can distill complex systems into real-time computational models that adhere to empirical observations and physical laws



Metocean data
(wind, waves, current)



Physics-informed reduced order models



Anchoring the future: strategic timeline

1. Benchmarking co-simulations

Benchmarking control stability during health-aware control against industry standards with injected noise and missing data to verify effectiveness and robustness under simulated conditions

2. Scaled hardware-in-the-loop (HiL)

Large-scale lab environments and demonstrator setups. This includes wind-tunnel testing of scaled (floating) offshore wind farms and drivetrain test benches.

3. Offshore shadow-mode

Deploying the integrated toolbox on live SCADA streams from operating turbines. The system generates control recommendations and RUL forecasts without actuating the turbine. Allows validation of the developed digital twins' performance with zero operational risk

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Questions?

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