Digital Twin Concept for Subsea Pipelines A Novel Framework

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AGENDA

Overview of Digital Twin # Application of Digital Twin in Offshore Applications # Case Study – Stress Prediction in Subsea Pipelines **#** Results & Discussions



Digital Twin - Introduction

- Digital twin technology is a virtual representation of a physical object, system, or process.
- It combines real-time data, simulation, and analytics to create a digital replica that can be used for monitoring, analysis, and optimization.
- In the context of offshore Industry, digital twin technology is used to create a virtual model of the Assets, capturing physical characteristics, performance data, and environmental conditions.



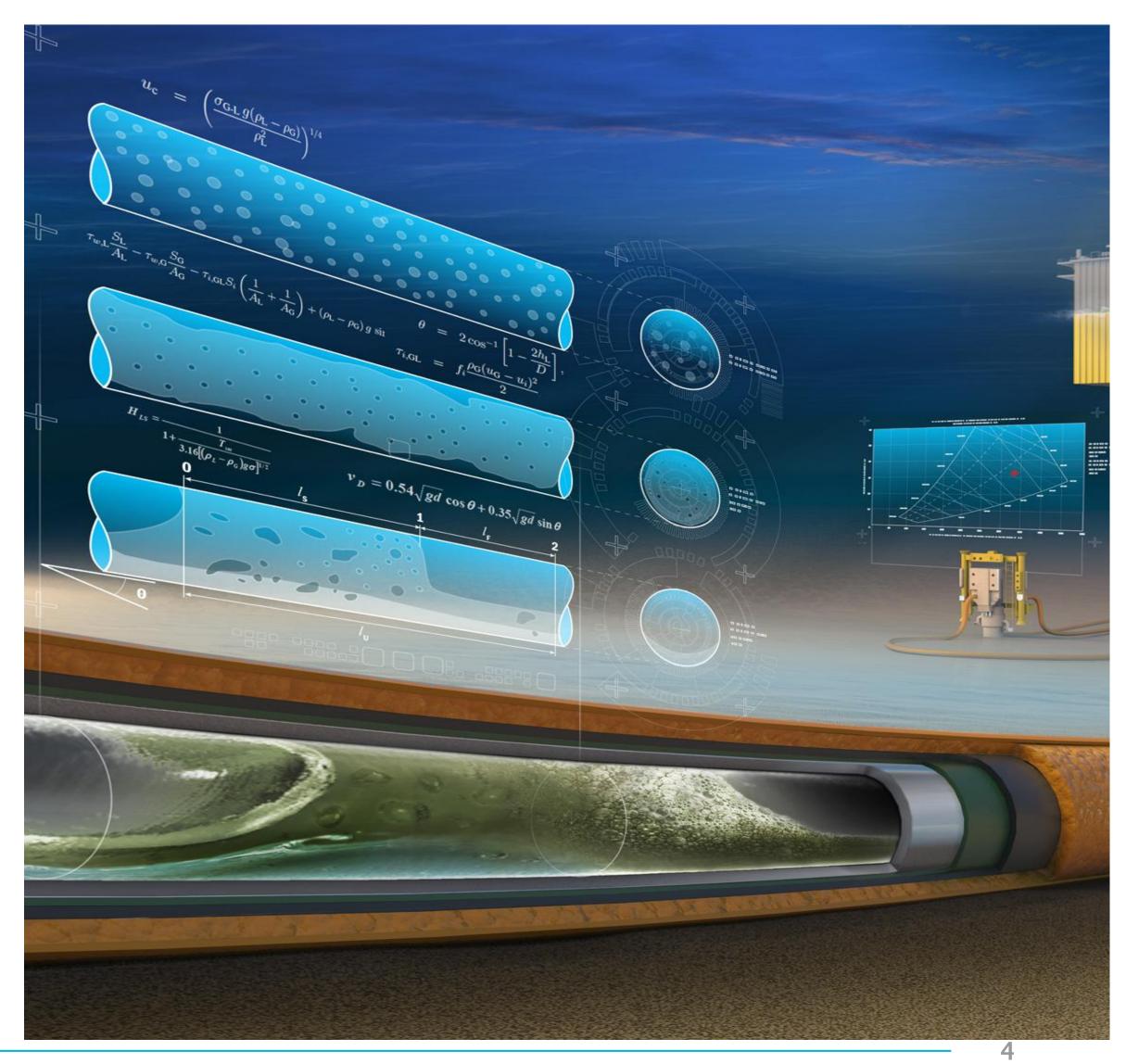




Digital Twin – How it Works

- Installation of Sensors
- •Data Collection and Transmission
- •Data Integration and Analysis
- •Creation of Digital Twin
- •Real-Time Monitoring and Predictive Maintenance.



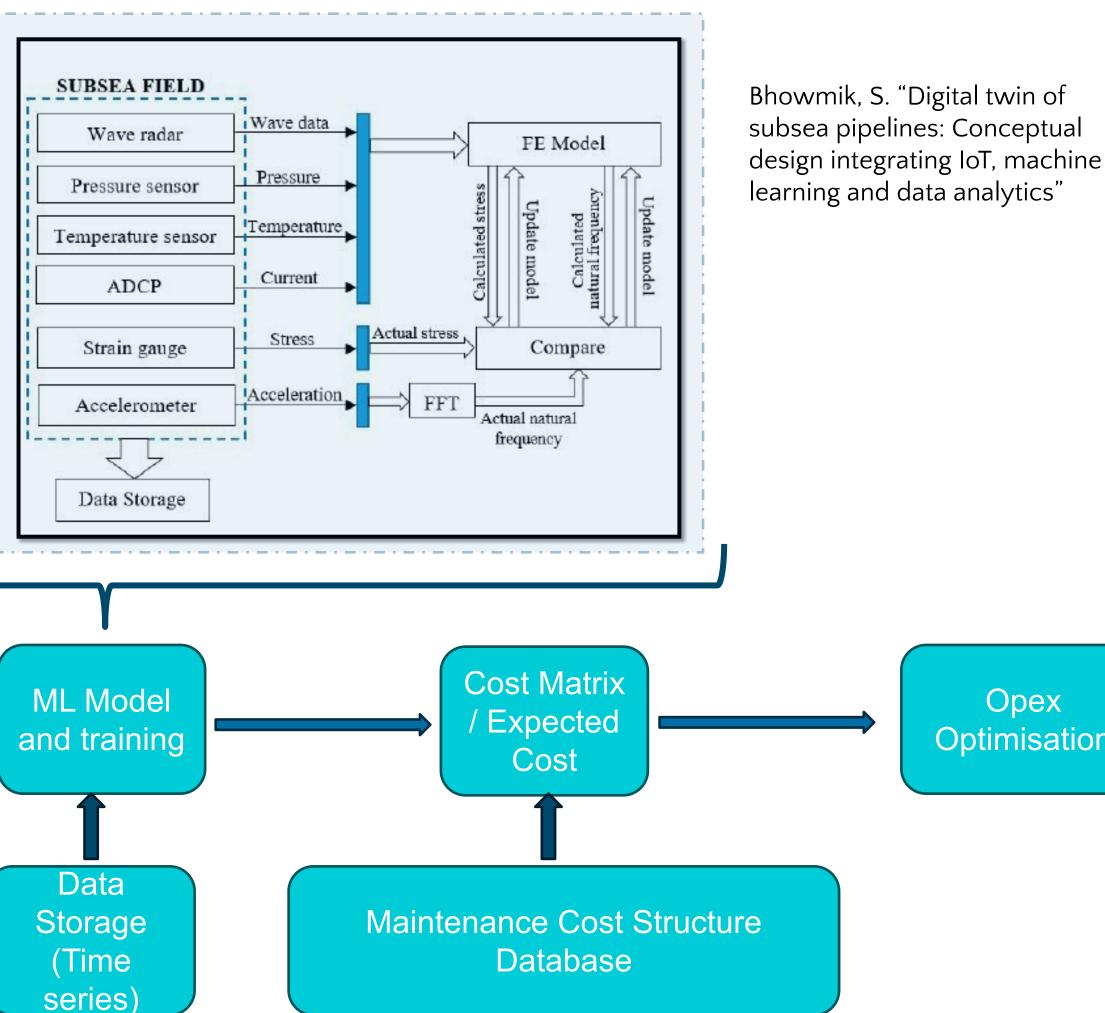




Data Integration and Analysis (e.g. Fatigue or **Predictive Maintenance)**

- One of the components is a computational model of the asset which is normally a finite element (FE) model
- The computational model of the subsea pipeline is updated based on different field sensor data
- Both the data-driven and FE-based models can be used to predict the remaining fatigue life
- Knowledge of the RFL can enable efficient maintenance planning and avoid unpredicted shutdowns



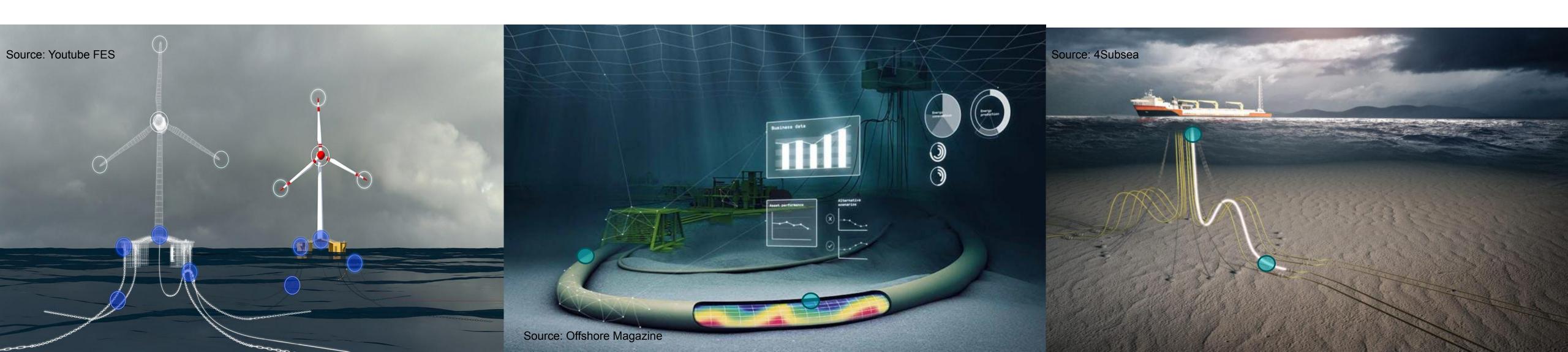






Challenges of Pre-trained Digital Twin Model

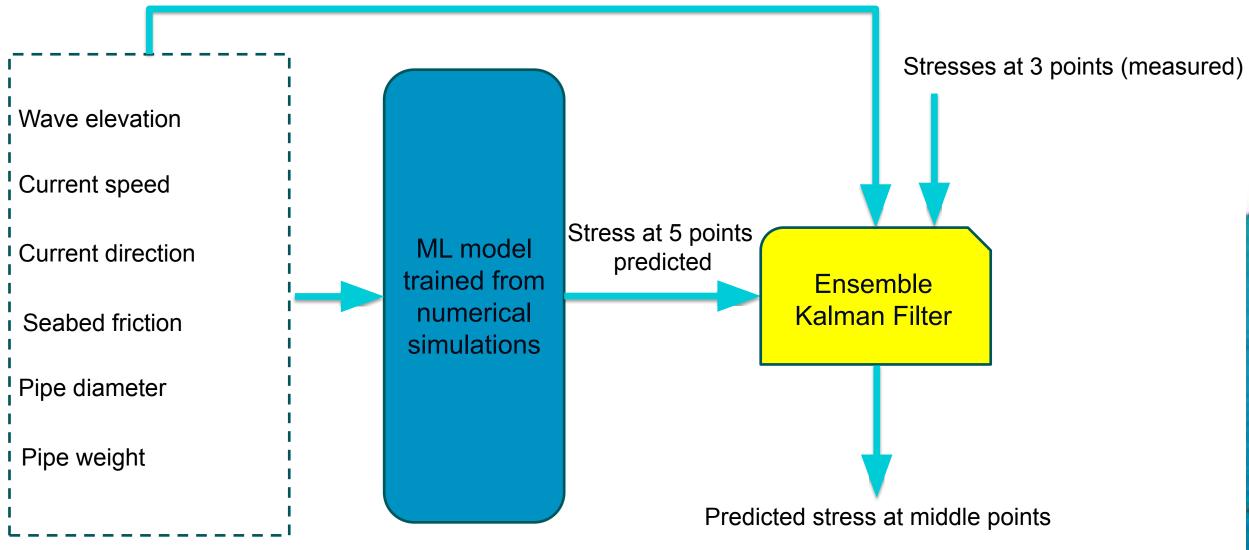
- Data acquisition at the right locations
- Quantity of sensors to equip the whole asset
- Change in environment post training is not captured
- Data transmission volume and speed





Novelty in Predicting Stresses

- Combining numerical simulations with measured data
- The physics driven ML model is built from numerical simulations
- The model is then calibrated and corrected online using
- The calibrated model is then used to estimate stress in I







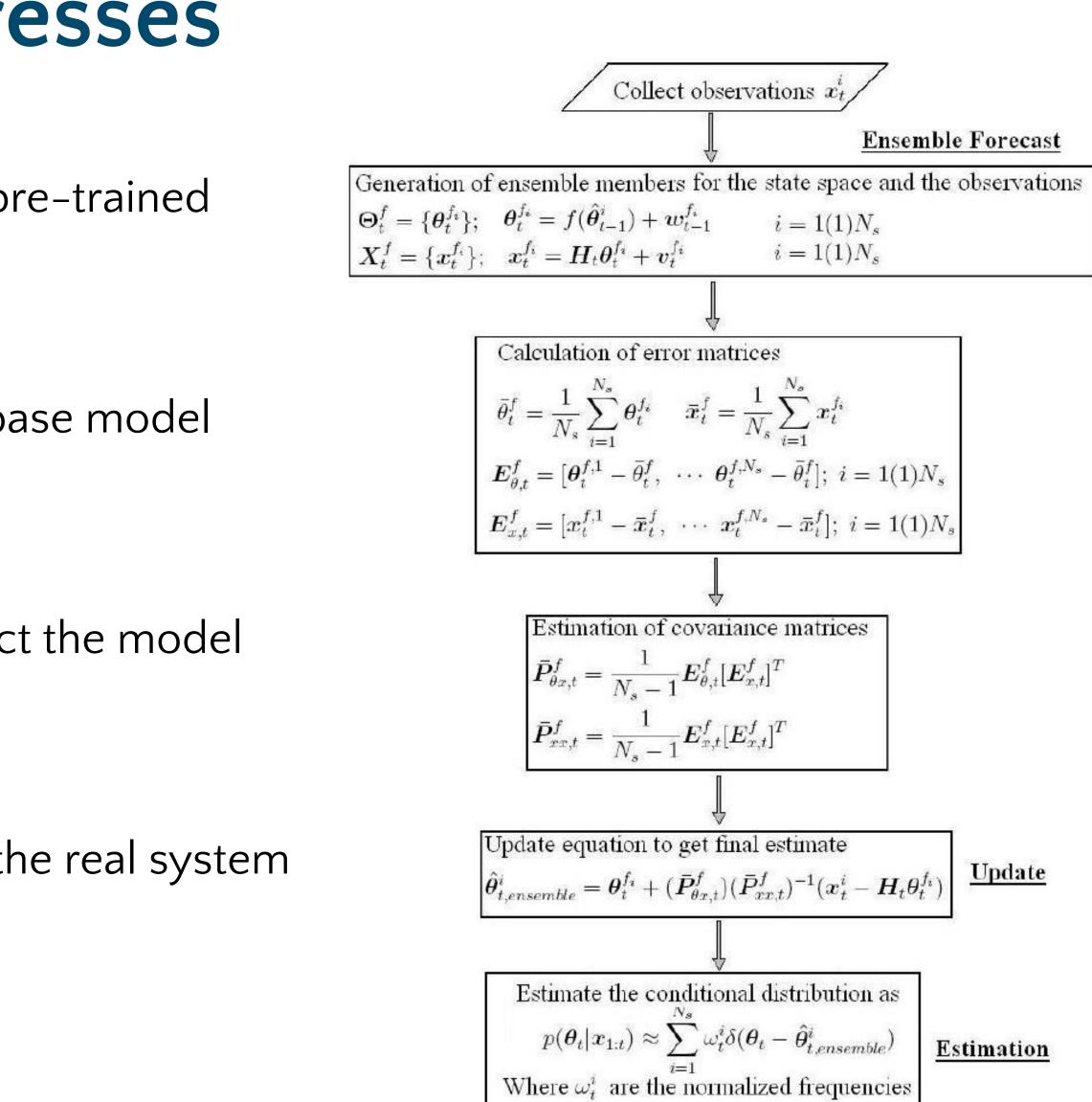


Novelty in Predicting Stresses Data assimilation

• The core idea is to correct the prediction of the pre-trained

model based on sparse measurements

- The pre-trained deep learning model acts as a base model representing the dynamic system
- Real-time measurements are then used to correct the model estimations of the system
- The differences are due to the imperfectness of the real system compared to the perfect model



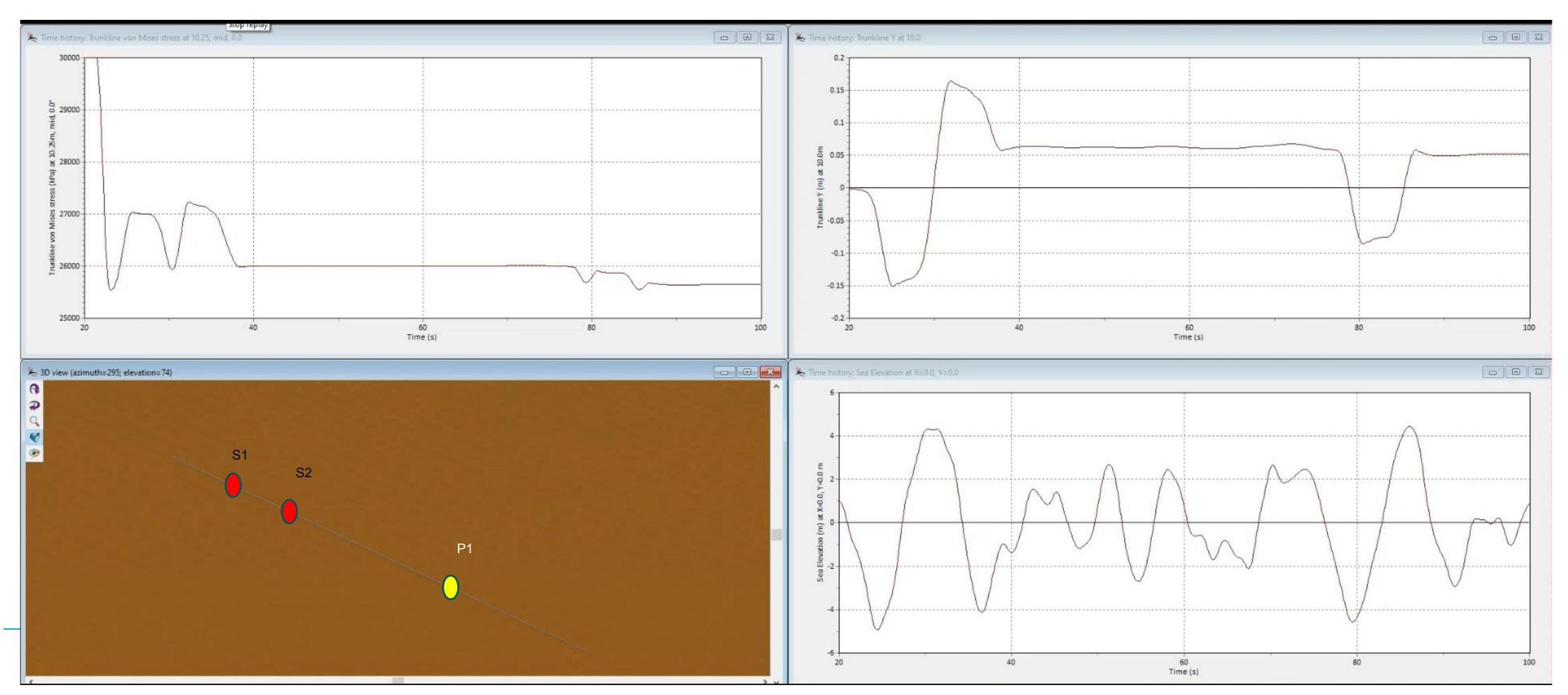
of the corresponding realizations



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Case Study – Subsea Pipeline

- Rigid steel pipe of 30in diameter 1 km long
- Water depth of 40m
- Waves of 7m Hs and 9 sec Tp, current of 1m/sec, applied in the perpendicular direction

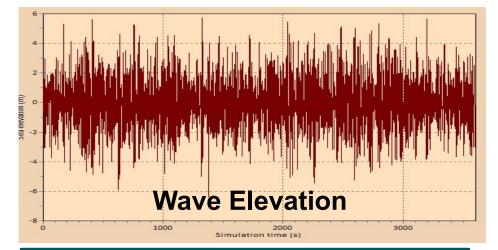




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Case Study – Stress Prediction

Input Data



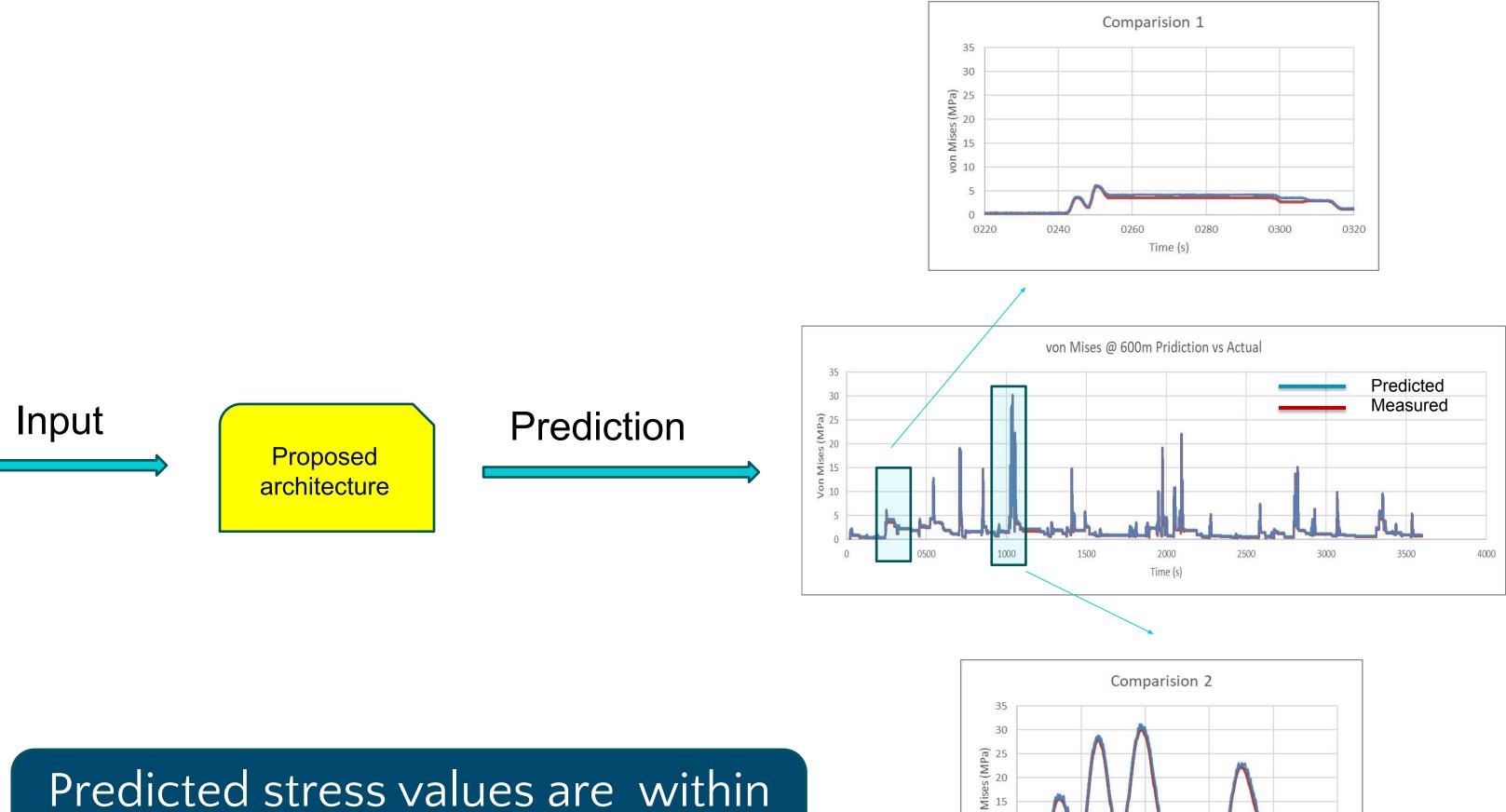
Current Speed, Block current 1 m/s

Current Direction, 90 deg to the pipeline axis

Seabed Friction 0.6

Pipe Diameter – 40"

Pipe dry weight – 890 kg/m



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Time (s)

Predicted stress 4% of the mea

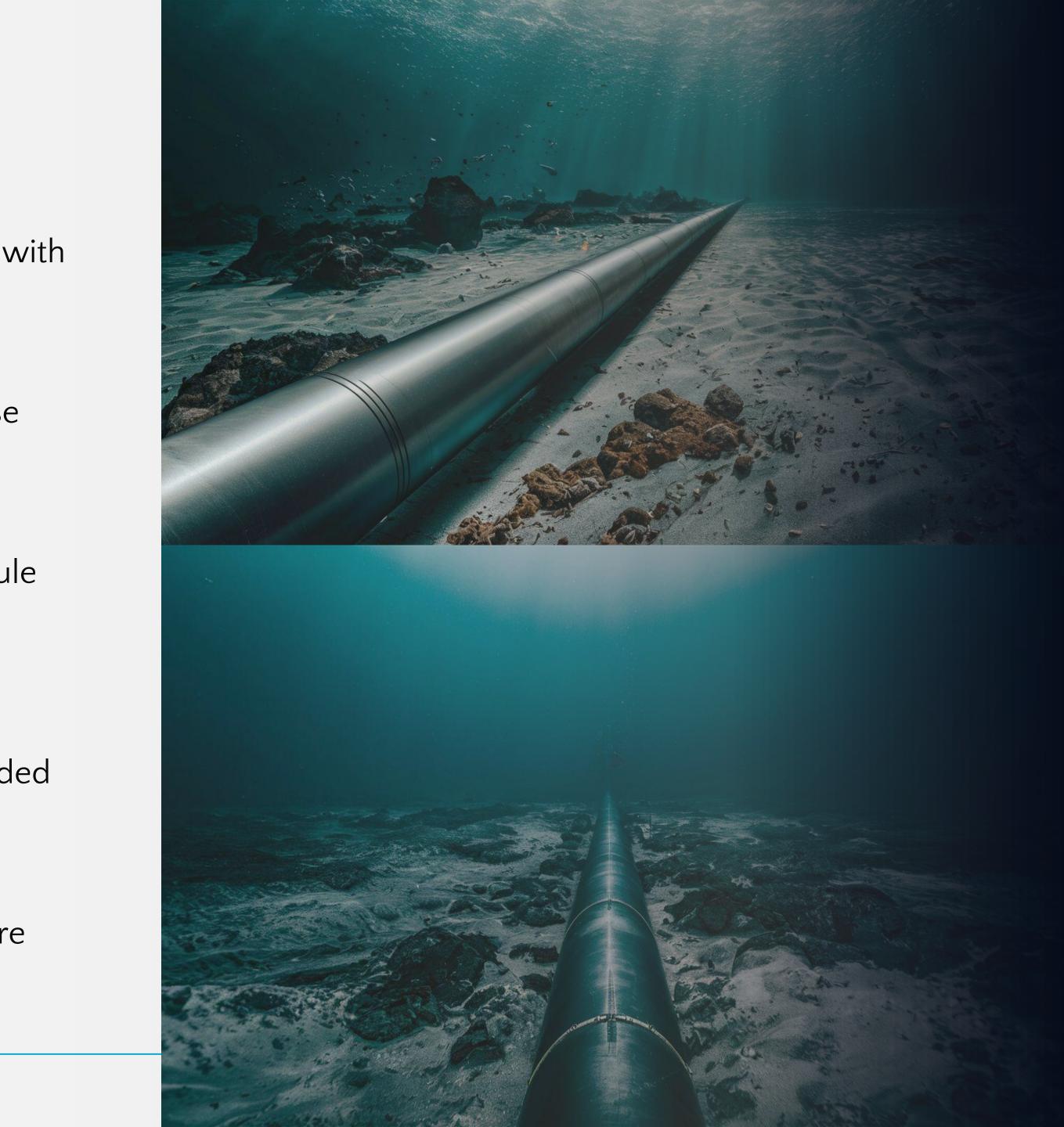


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4% of the measured stresses.

Summary

- A novel approach of combining numerical simulations with measured data is presented to predict stresses in real-time.
- Accuracy greater than 95 % is achieved on the test case presented for a subsea pipeline
- Real-time monitoring helps in the reduction of ROV inspection and aids the predictive maintenance schedule
- Tracking stress and fatigue improves useful remaining fatigue life using actual loading condition
- The framework presented is generic and can be extended to various subsea applications
- The system relies on the accuracy of sensor data and transmission interval which is a challenge as we venture into deeper waters



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Questions...