

Instrumenting the Ocean

Data Capabilities of HarshLab

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● Bilbo
● 27/05/2026





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1. Introduction

2. Sensors

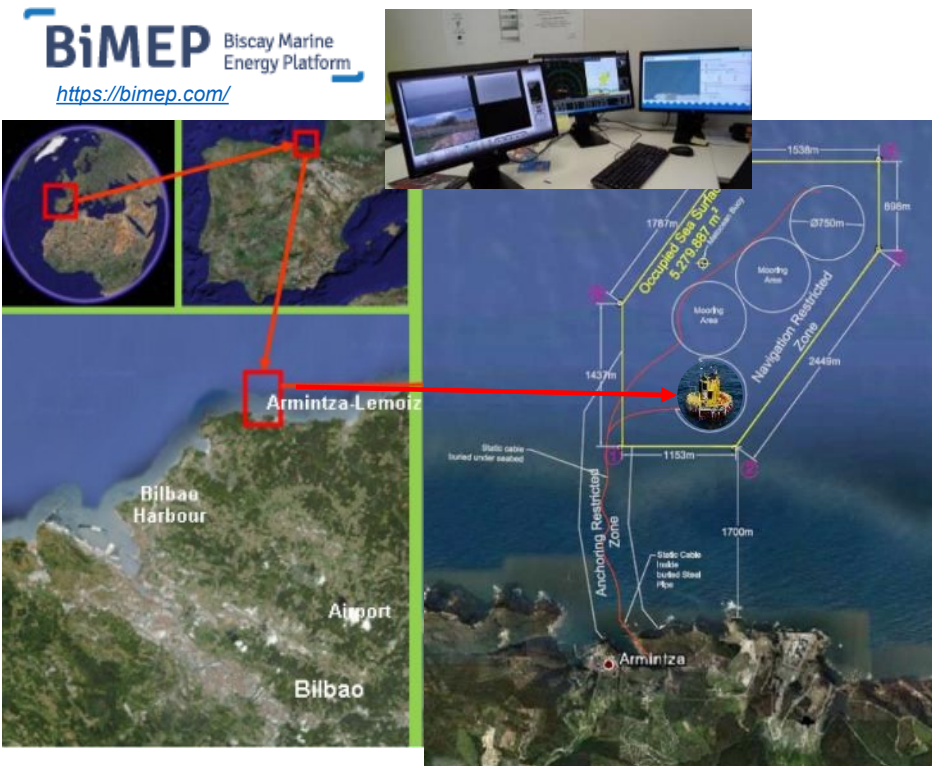
3. Why implement sensors?

4. Studies based on HarshLab data: RUL-ET project



1. Introduction

What is HarshLab and where is it located?



- **Floating laboratory**, moored at Bimep, Arminza (Bizkaia).
- At a depth of 65 meters and 1.6 nautical miles from the coast.
- BiMEP is an open-sea test area with grid connection for the demonstration and validation of ocean energy technologies.
- Fully equipped with subsea infrastructure for connection to the onshore grid::
 - **4 subsea cables 13.2 kV – 5 MW**
 - Controlled and monitored environment
 - **24/7 surveillance**
- Quick access from Arminza port (10 min)



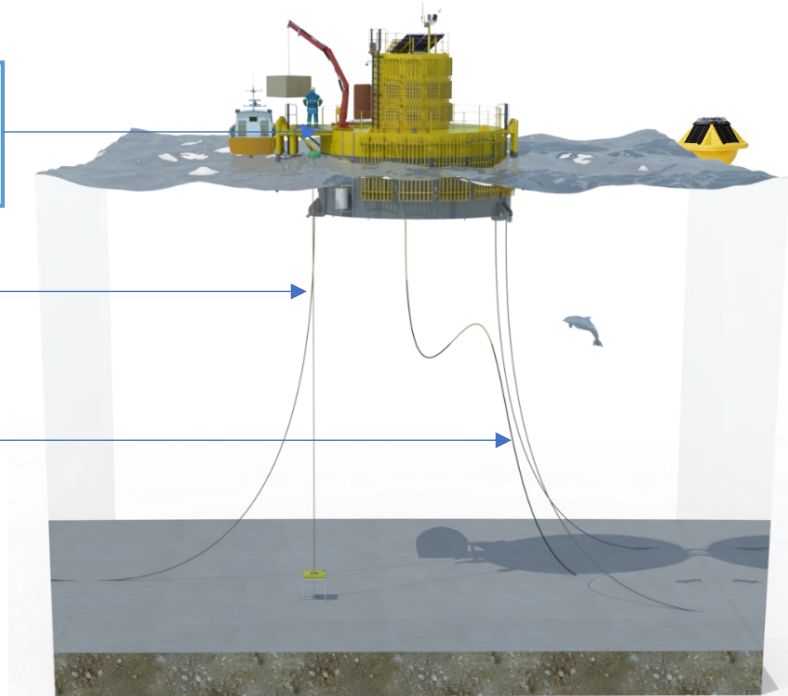
What can be tested?

- **Testing of materials and components** in a real offshore environment: **atmospheric, splash, immersion, confined areas, and seabed**
- Testing of solutions for **protection** against **corrosion, fouling, and corrosion**, fatigue.
- Monitoring and **data acquisition** for AI-ML models for **O&M** and design. Example: Structural, mooring lines, and dynamic cable.
- **Monitoring of environmental** conditions and **marine fauna**
- **Training personnel** in offshore operations

<https://harshlab.eu/>

3. Sensors

What can be monitored in HarshLab?



2. Floating structure
(hull, equipment, corrosion,
materials)

3. Mooring lines, anchors

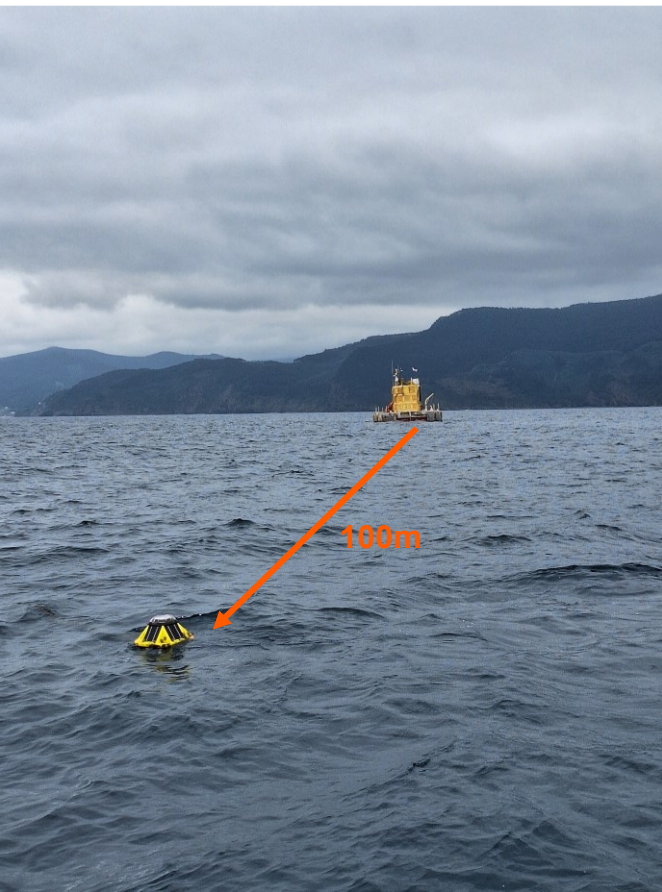
4. Dynamic Cable,
Connector, bend stiffener,
buoyancy modules

1. Environmental and
Metocean conditions


Marine fauna (Birds, fish...)

Navigation

What are we monitoring in HarshLab?



1. Metocean and environmental conditions: Spotter and weather station

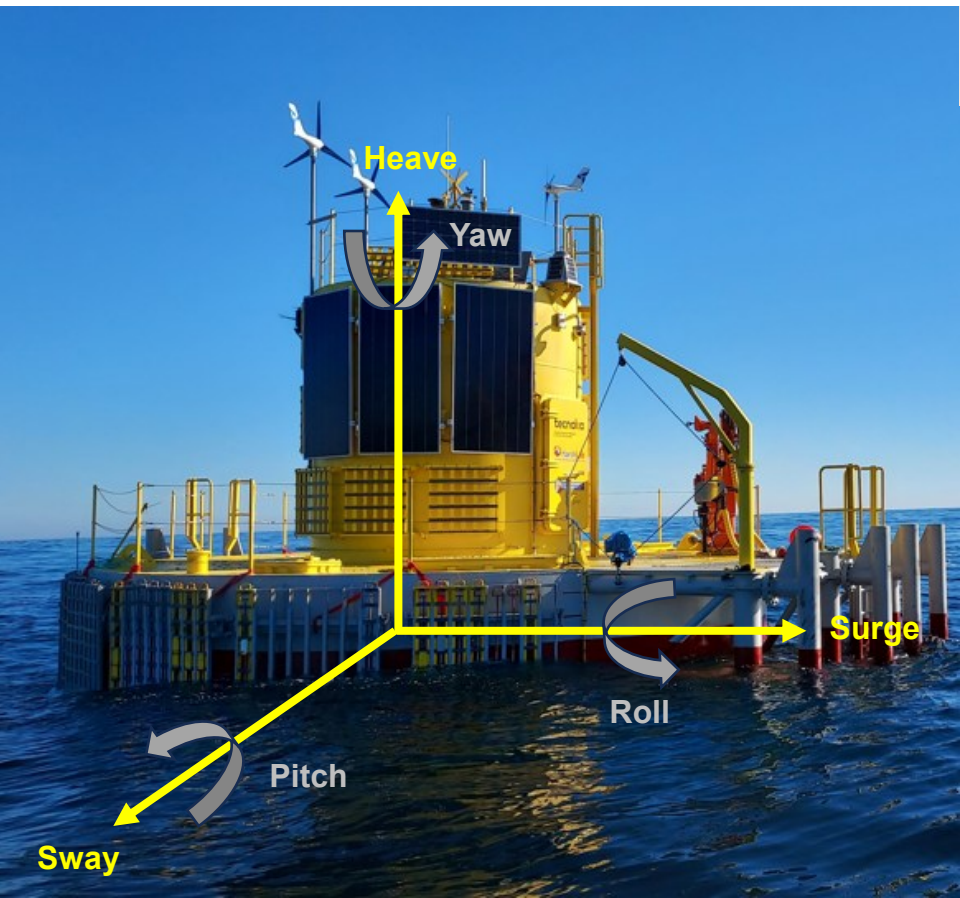
 **Spotter Metocean buoy** : $H_s (m)$, $T_p (s)$, directional spectrum, air Pressure, Wind (speed and direction), Water temperature..

 **Meteorological data**: Air Temperature and pressure, Humidity, Wind (speed and direction), Rain rate, Solar radiation

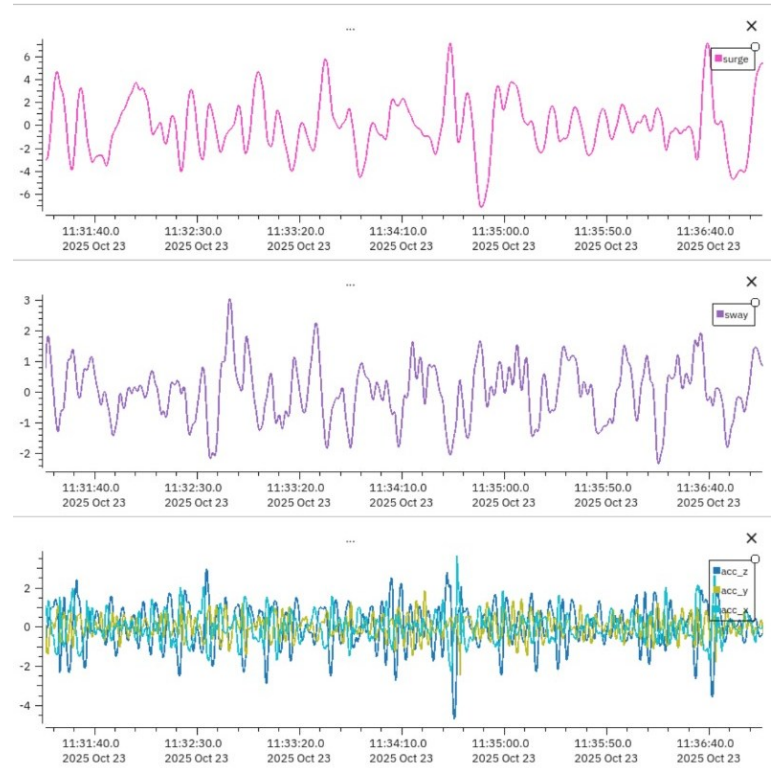
 **Underwater environmental data**: Hydrophone for detection of marine fauna (future)

 **Underwater camera** : Fish detection

What are we monitoring in HarshLab?

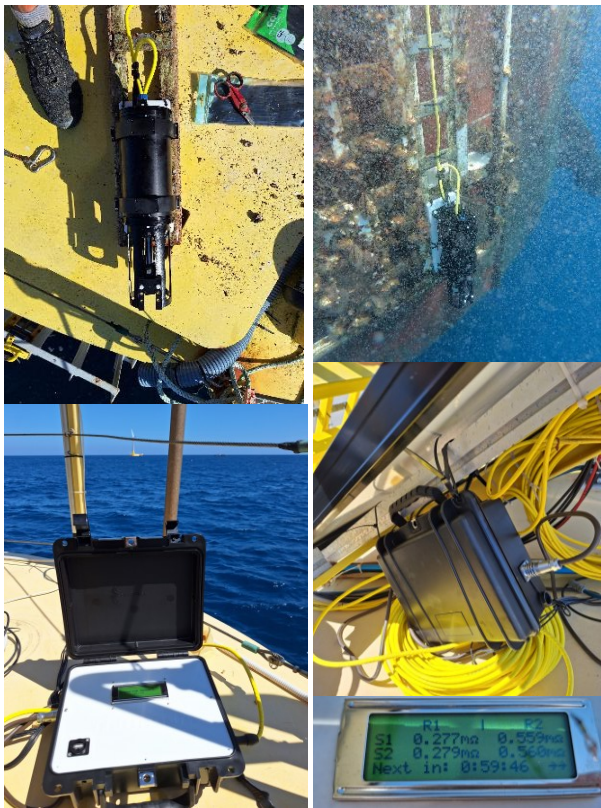


2. Floating Structure 6 Degrees of freedom (GNSS, IMU)



What are we monitoring in HarshLab?

2. Materials: Corrosion Sensor (joint development Vicinay–Tecnalia)



📡 Determine the corrosion rate in structures submerged in the marine environment

📡 Estimate the **remaining life** of the element

How does it work?

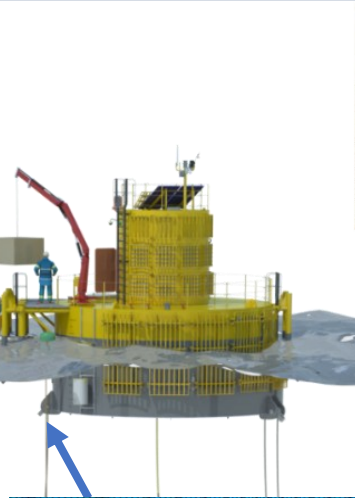
- An electric current is passed through a specimen made of the same material as the structure to be monitored.
- The electrical resistance of the specimen is measured.
- The variation in resistance indicates material loss due to corrosion.
- The corrosion progress in mm/year is calculated from that resistance difference.

📡 Current status: **TRL5 at HarshLab**, patent PC25-00866-148

What are we monitoring in HarshLab?



3. Mooring System: tensions

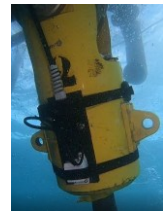
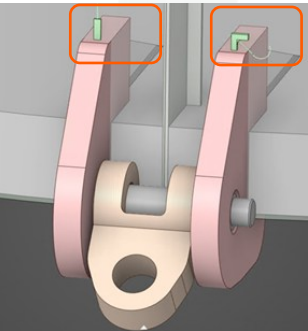
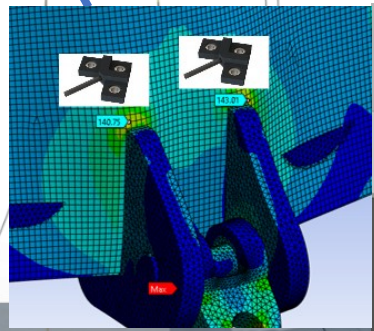


✓ **Strain Gauges (indirect measurement from the inside)**
12 strain gauges inside the hull (4 per mooring point) in areas of maximum mechanical stress..

✓ **Inclinometers for Mooring Lines + Receiver Hydrophone**
(Tecnalia-CoreMarine – INTEGRIA project, Float&OM)

✓ **Strain Gauges (indirect measurement from the outside)**
2 strain gauges on top of each of the 3 padeyes, on the outside of the hull – **ONGOING – Gauges purchased, to be installed**

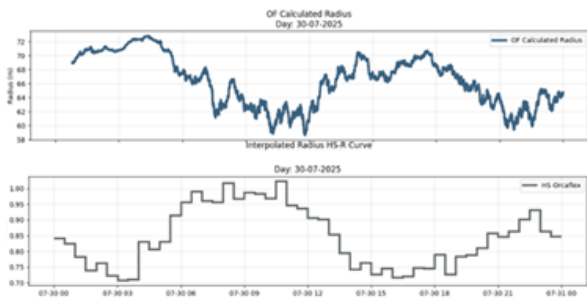
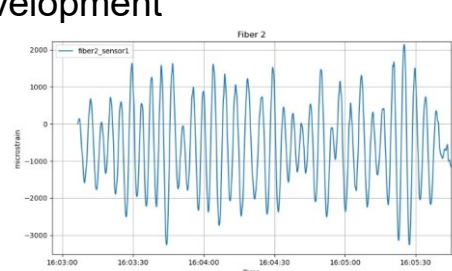
✗ Direct measurement strain gauge for mooring line tension



What are we monitoring in HarshLab?

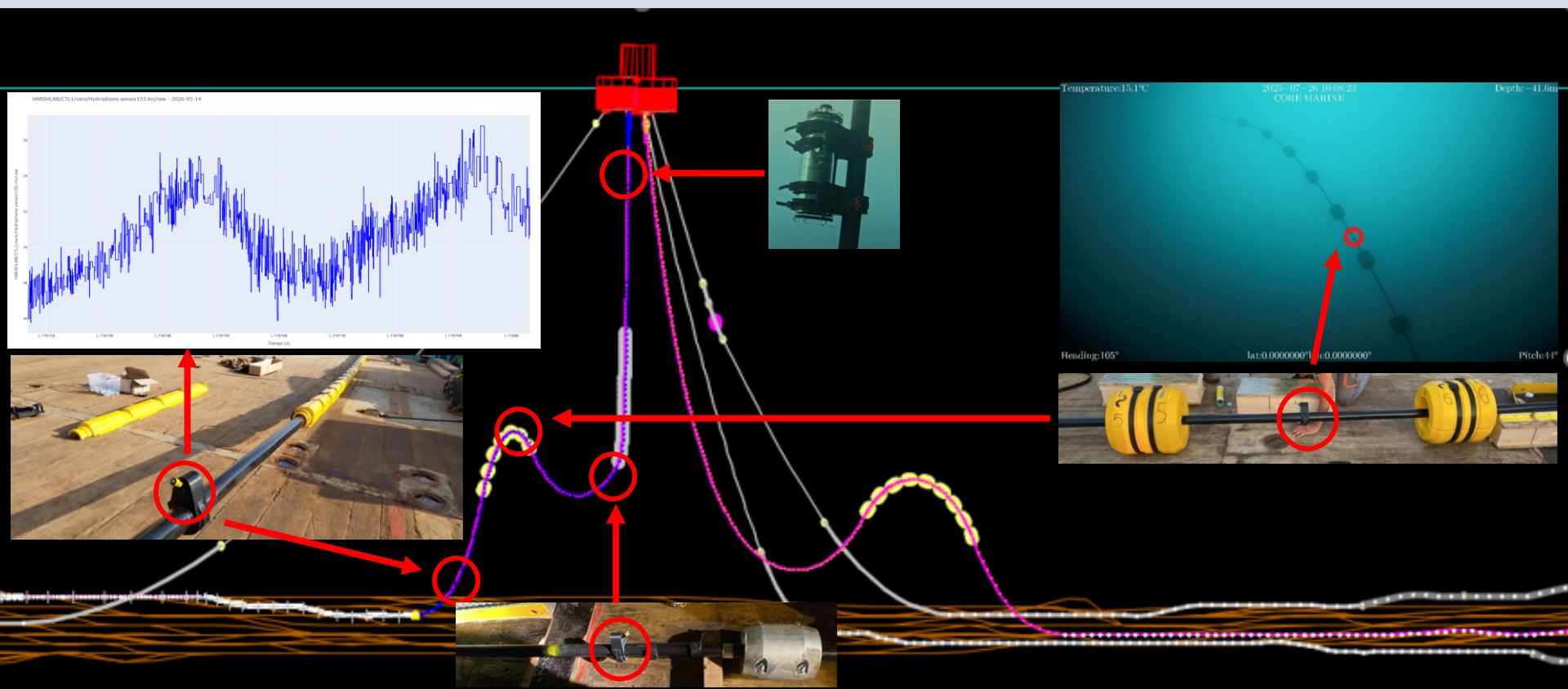
4. Dynamic Cable: Tensions and curvatures

- ✓ **Bend Stiffener** – 4 strain gauges (curvature measurement)
- ✓ **Bend Stiffener** 4 fiber optic sensors for curvature measurement. Inialia development





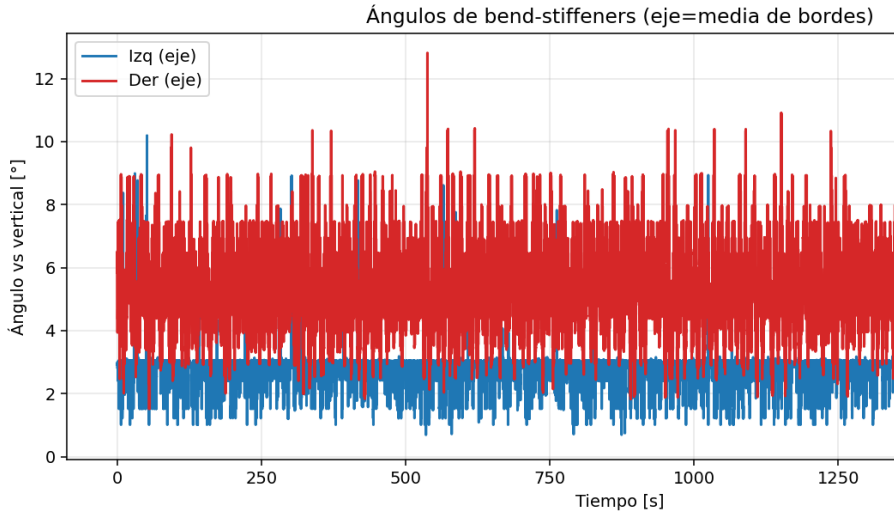
4. Dinamic cable: Inclinations at different points (CoreMarine –Tecnalia)



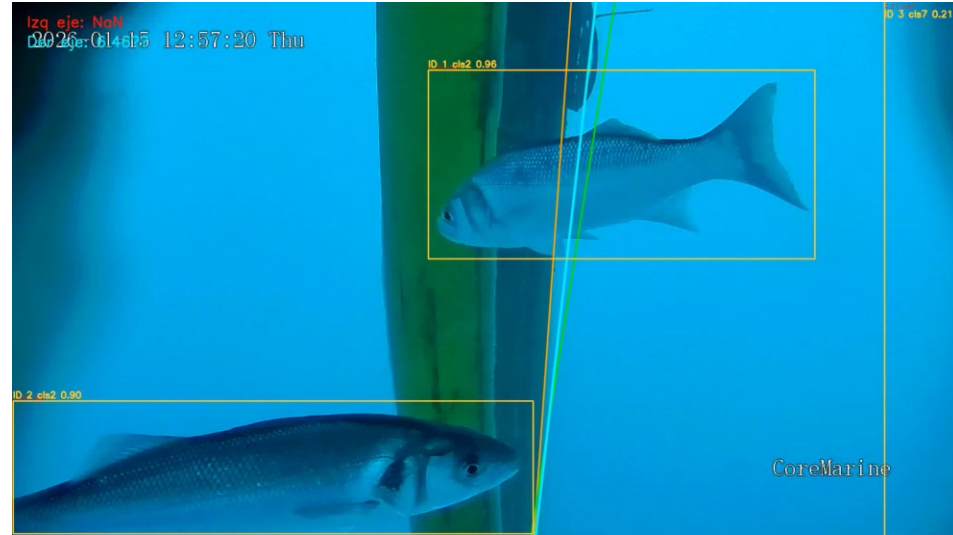
What are we monitoring in HarshLab?

4. Dynamic cable: Monitoring Bend Stiffener and fish with underwater camera

✓ Classical Computer vision for angle measurement



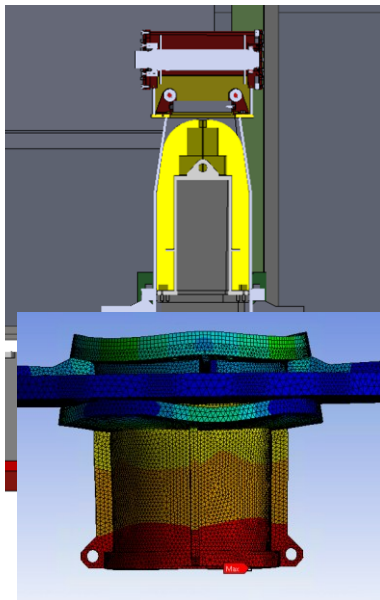
✓ Fish detection Based on YOLOv8 trained model



What are we monitoring in HarshLab?

4. Dynamic Cable-Ditrel Connector: Stresses

- ✔ **Strain gauges** installed on the **connector** for axial load measurement



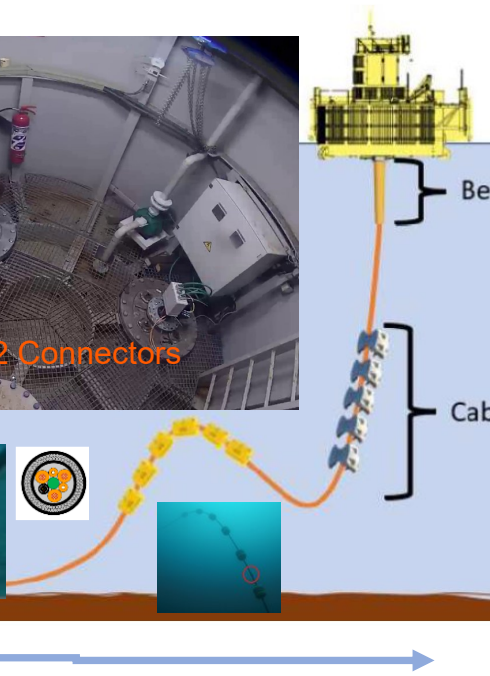
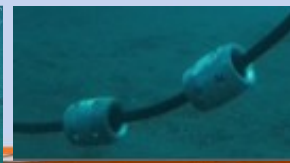
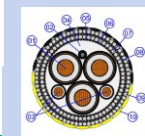
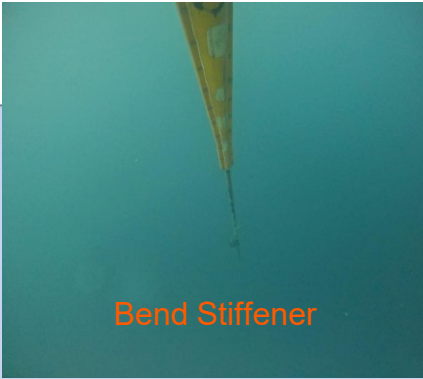
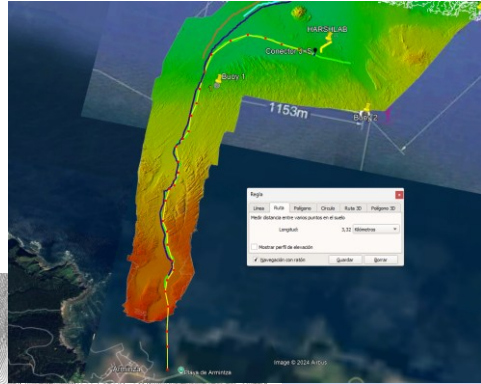
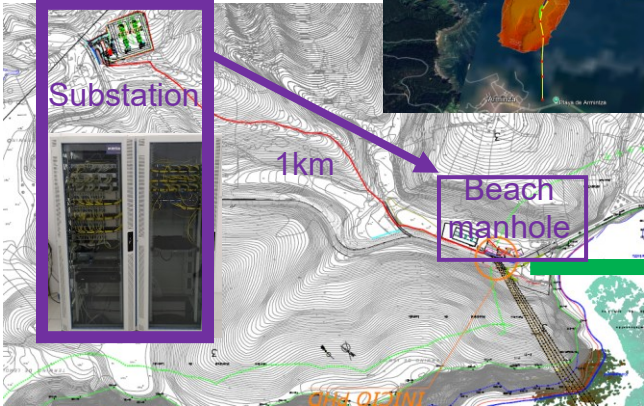
What can be monitored in HarshLab?

4. Dynamic Cable: Possibility of monitoring the dynamic cable through fiber optics

✓ European project proposals submitted in February

Bimep Office

24 optic fiber monomode
Wave-length 1310-1550



Static Cable (3.5km)
3x185 mm² + 2 x 70 mm² 0.6/1 kV + 24FO

Dynamic Cable (280m)
3 x 50 mm² + 2 x 6 mm² 0.6/1 kV + 24FO (OD = 58.5 mm)

HarshLab as Data Source → DasWind

Bienvenido Admin



Cpu: 0%



13.95% Ocupado



28 Alarmas



4G (LTE)



-66 dBm



Up: 6d 18h 41m



0 V



Core Running

Sistemas de adquisición

Core



✓ Online

Quantum



✓ Online

Sistema Inercial



✓ Online

Davis Weather Station



✓ Online

Twincat ADS



✓ Online

Sistema Inercial CM



✓ Online

Hydrophone CM



✓ Online

GPS CM



✓ Online

Sistema Inercial TP



▲ Offline

Twincat ADS TP



⌚ Lost

Router



✓ Online

DAS

DAS HarshLab

Offshore Materials & Components Lab
HarshLab



Sesion

Admin

Privilegios: A Console

Ultima sesion: 25/05/2026 (134.231.183.121)



Conexion movil

4G (LTE): N/A

IP/Gateway: 88.28.223.226 / 88.28.223.226

Signal Level: -66 dBm

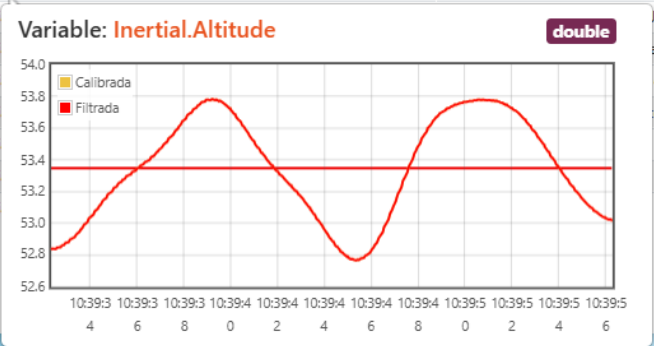


Variables

[Todo](#)
[System](#)
[Inertial](#)
[HBM](#)
[Twincat](#)
[Env](#)
[Meteo](#)
[Tracker](#)
[TP](#)
[Gauge](#)
[test](#)

Mostrar: líneas

Variable	Descripcion	Category	Value
Inertial.VarStatus	Estado de las variables de la IMU-GPS	Inertial	0 NOARG
Inertial.Euler.Yaw	Angulo de Euler YAW	Inertial	57.863058 deg
Inertial.Euler.Pitch	Angulo de Euler PITCH	Inertial	1.455314 deg
Inertial.Euler.Roll	Angulo de Euler ROLL	Inertial	-5.546236 deg
Inertial.Acc.X	Aceleración en el eje Longitudinal de la Nacelle	Inertial	0.162 m/s2
Inertial.Acc.Y	Aceleración en el eje transversal de la Nacelle	Inertial	-0.65 m/s2
Inertial.Acc.Z	Aceleración en el eje Z de la Nacelle	Inertial	-9.743 m/s2
Inertial.Gyro.X	Angular Rate X axis	Inertial	-0.234913 deg/sec
Inertial.Gyro.Y	Angular Rate Y axis	Inertial	0.641713 deg/sec
Inertial.Gyro.Z	Angular Rate Z axis	Inertial	-0.085944 deg/sec
Inertial.Latitude	Latitude	Inertial	43.456283 deg
Inertial.Longitude	Longitude	Inertial	-2.884973 deg
Inertial.Altitude	Altitude	Inertial	53.556 m





Datos en directo

[Portal](#) / [Variables](#) / [Listado](#)

Variables

[Todo](#) [System](#) [Inertial](#) **[HBM](#)** [Twincat](#) [Env](#) [Meteo](#) [Tracker](#) [TP](#) [Gauge](#) [test](#)

Mostrar: líneas

Variable	Descripcion	Category	Value
Hbm.Signal1	Canal 1 Quantum	HBM	-2.074903 -
Hbm.Signal2	Canal 2 Quantum	HBM	-4.386075 -
Hbm.Signal3	Canal 3 Quantum	HBM	1999999973982208 -
Hbm.Signal4	Canal 4 Quantum	HBM	7999999895928832 -
Hbm.Signal5	Canal 5 Quantum	HBM	7999999895928832 mV/V



Datos en directo

Portal / Variables / Listado

Variables

Todo System Inertial HBM Twincat Env **Meteo** Tracker TP Gauge test

Mostrar: 50 líneas

Variable	Descripción	Category	Value
Meteo.Temperature	Temperature	Meteo	21.4 C
Meteo.Humidity	Humidity	Meteo	79 %
Meteo.Pressure	Barometric pressure	Meteo	1024.3 mbar
Meteo.AbsPressure	Absolute barometric pressure	Meteo	1024.3 mbar
Meteo.WindSpeed	Wind speed	Meteo	16 kph
Meteo.WindDirection	Wind direction	Meteo	59 deg
Meteo.RainRate	Rain rate	Meteo	0 mm/hour
Meteo.UVIndex	UV Index	Meteo	51 -
Meteo.SolarRadiation	Solar radiation	Meteo	610 W/m2

Datos en directo


















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Variables

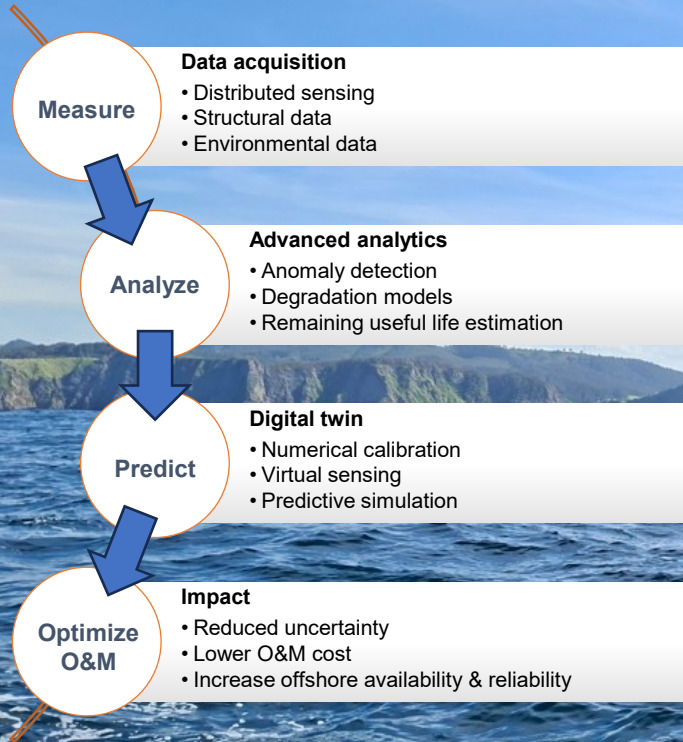
[Todo](#)
[System](#)
[Inertial](#)
[HBM](#)
[Twincat](#)
[Env](#)
[Meteo](#)
[Tracker](#)
[TP](#)
[Gauge](#)
[test](#)

Mostrar: líneas

Variable	Descripción	Category	Value
Tracker.Gps.alt	Altitud	Tracker	3.3188 m
Tracker.Gps.er_alt	Auto	Tracker	0.035 -
Tracker.Gps.er_lat	Auto	Tracker	0.015 -
Tracker.Gps.er_lon	Auto	Tracker	0.013 -
Tracker.Gps.fix	Adquisicion valida	Tracker	4 -
Tracker.Gps.hdp	Auto	Tracker	0.7 -
Tracker.Gps.lat	Latitud	Tracker	43.456276 deg
Tracker.Gps.lon	Latitud	Tracker	-2.885015 deg
Tracker.Gps.pdop	Auto	Tracker	1.4 -
Tracker.Gps.sats	Numero de satelites	Tracker	16 -
Tracker.Gps.vdop	Auto	Tracker	1.2 -
Tracker.Inertial.Euler.Yaw	Angulo de Euler YAW	Tracker	-136.7601 deg
Tracker.Inertial.Euler.Pitch	Angulo de Euler PITCH	Tracker	1.2887 deg
Tracker.Inertial.Euler.Roll	Angulo de Euler ROLL	Tracker	-0.508 deg
Tracker.Inertial.Acc.X	Aceleración en el eje Longitudinal de la Nacelle	Tracker	0.303 m/s2
Tracker.Inertial.Acc.Y	Aceleración en el eje transversal de la Nacelle	Tracker	0.1047 m/s2
Tracker.Inertial.Acc.Z	Aceleración en el eje Z de la Nacelle	Tracker	-9.7561 m/s2
Tracker.Inertial.Temperature	Temperature	Tracker	36.65 °C
Tracker.Inertial.Heave	Desplazamiento vertical respecto a posición media	Tracker	0.0347 m
Tracker.Inertial.HeaveRate	Velocidad vertical	Tracker	-0.0632 m/s
Tracker.Inertial.DelayedRate	Heave corregido	Tracker	-0.3858 m

Hydrophone.line	Ultima linea NMEA	Tracker	0 --	2.0 s
Hydrophone.lines	auto	Tracker	36443 --	2.0 s
Hydrophone.sensor104.acc	 Aceleración RMS	Tracker	43 m/s2	3m 34s
Hydrophone.sensor104.inc	 Inclinación	Tracker	19 deg	3m 34s
Hydrophone.sensor105.acc	 Aceleración RMS	Tracker	14 m/s2	7m 29s
Hydrophone.sensor105.prof	 Profundidad	Tracker	10 m	7m 29s
Hydrophone.sensor106.acc	 Aceleración RMS	Tracker	0 m/s2	
Hydrophone.sensor106.inc	 Inclinación	Tracker	0 deg	
Hydrophone.sensor107.acc	 Aceleración RMS	Tracker	0 m/s2	
Hydrophone.sensor107.prof	 Profundidad	Tracker	0 m	
Hydrophone.sensor110.acc	 Aceleración RMS	Tracker	40 m/s2	4m 54s
Hydrophone.sensor110.inc	 Inclinación	Tracker	16 deg	4m 54s
Hydrophone.sensor111.acc	 Aceleración RMS	Tracker	16 m/s2	8m 50s
Hydrophone.sensor111.prof	 Profundidad	Tracker	7.5 m	8m 50s
Hydrophone.sensorTBR.noiseMean	 Ruido medio	Tracker	9 --	4m 5s
Hydrophone.sensorTBR.noisePeak	 Ruido pico	Tracker	15 --	4m 5s
Hydrophone.sensorTBR.temp	 Temperatura	Tracker	23.200001 °C	4m 5s
Hydrophone.sensor153.inc	Inclinación	Tracker	47.9 deg	2.0 s
Hydrophone.sensor153.std	Desviación típica	Tracker	0 deg	2.0 s
Hydrophone.sensor154.inc	 Inclinación	Tracker	0 deg	
Hydrophone.sensor154.std	 Desviación típica	Tracker	0 deg	
Hydrophone.sensor155.inc	Inclinación	Tracker	52 deg	8.0 s
Hydrophone.sensor155.prof	Profundidad	Tracker	23 m	8.0 s

Why monitor offshore floating structures?



4. Studies based on HarshLab data

RULET project



Motivation and project approach

STUDY A:  ASSESSMENT OF MOORING LINE DEGRADATION 

STUDY B:  UMBILICAL CABLE DAMAGE MONITORING 

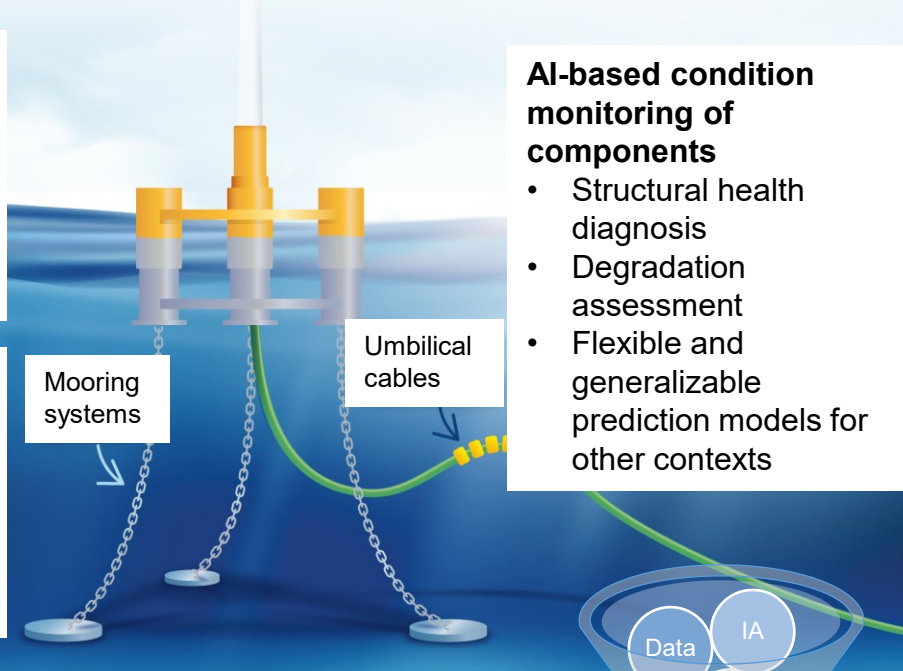
OBJECTIVE: Estimate the Remaining Useful Life of offshore mooring systems and dynamic cables

Efficient dataset generation strategies

- Acceleration of physical testing in the laboratory
- Failure data simulation

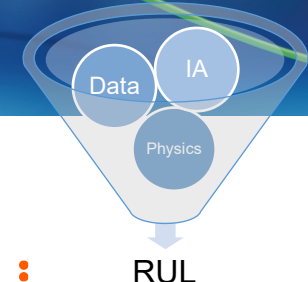
Decision-making tools for predictive maintenance

- Multi-objective optimisation
- Uncertainty management



AI-based condition monitoring of components

- Structural health diagnosis
- Degradation assessment
- Flexible and generalizable prediction models for other contexts



Motivation



Increasing unit power



Increasing structural loads



High access and maintenance costs



Energy production losses in case of failure

AI is promising - but DATA is essential

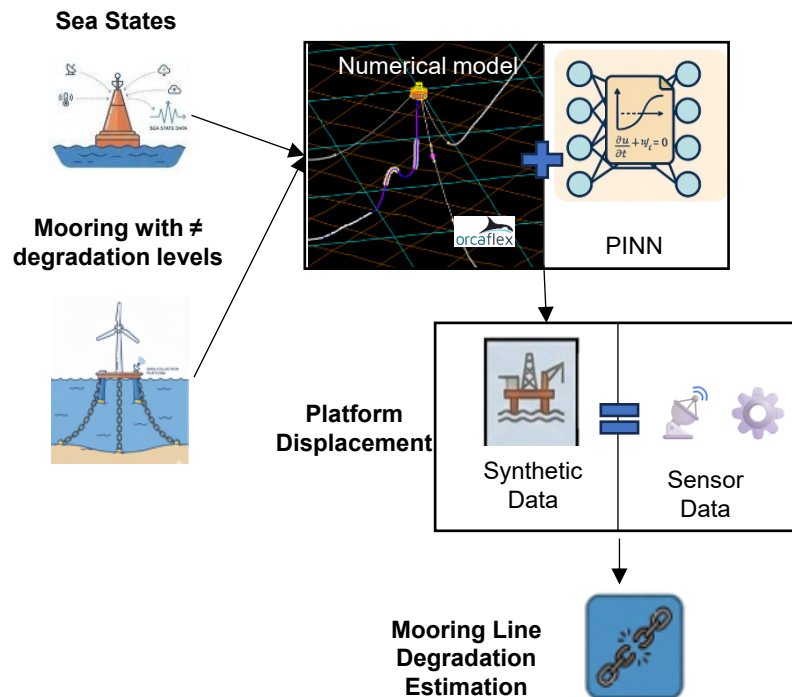
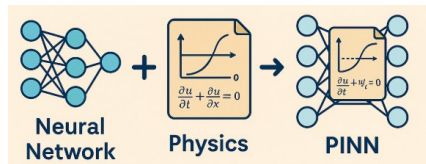
Decisions are economically driven → RELIABILITY is critical

STUDY A: ASSESSMENT OF MOORING LINE DEGRADATION (I)

- **Objective:** To estimate corrosion based on variations in floater motions
- **Approach:** direct measurement of corrosion is challenging
 - numerical model → synthetic data generation
 - PINN trained across:
 - sea states and
 - degradation levels

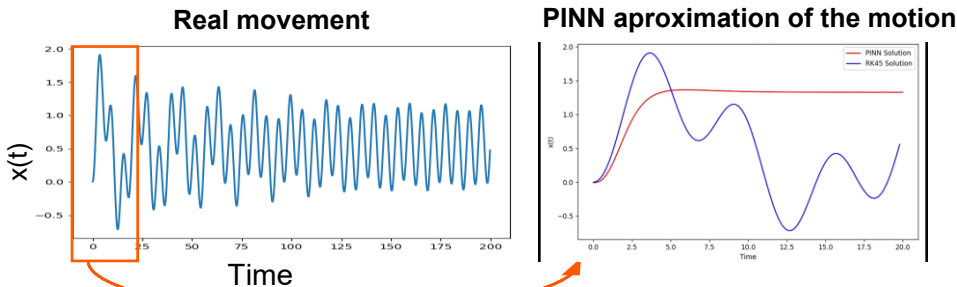
Why Physics-Informed Neural Networks-PINNs??

- Combine physics and data
- Improving accuracy and consistency
- Less data required
- Better generalization
- Fast simulations



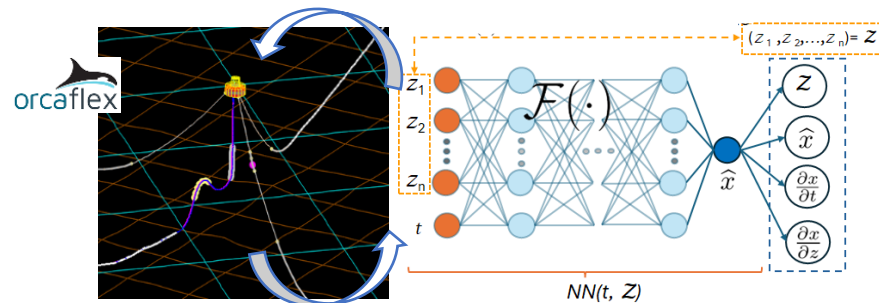
STUDY A: ASSESSMENT OF MOORING LINE DEGRADATION (II)

Training 1: simplified physics (single DOF – surge)



PINNs learn the physics, not just the data and mimic the physics-based solver

Training 2: PINN connected to OrcaFlex (6 DOF)



Automatized simulation:
PINNs ↔ Numerical model

Key findings

- Limited impact of sea state on accuracy
- Degradation level drives prediction error
- Best performance at intermediate degradation levels

In moorings lines, what degrades matters more than the sea state

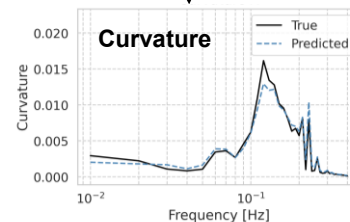
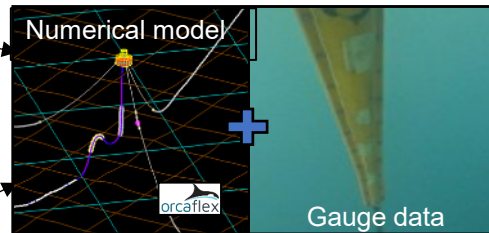
STUDY B: UMBILICAL CABLE DAMAGE MONITORING (I)

- **Objective:** To predict curvature PSD at a given location
- **Inputs:**
 - Metocean data (Hs, Tp, direction, currents)
 - Platform motion (6 DOF response spectra)
- **Output:** Curvature PSD → damage estimation
- **Approach:**
 - Machine learning surrogate model
 - Trained with synthetic and measured data

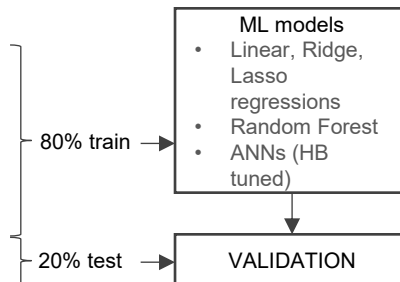
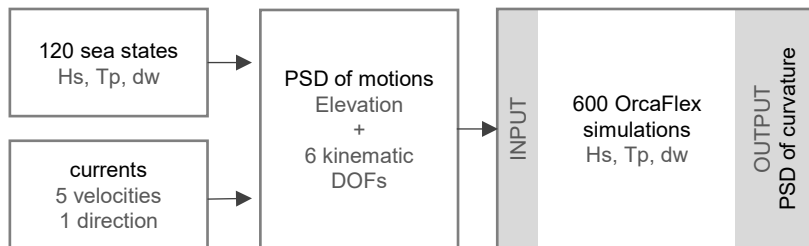
Sea States



Platform motion



Accumulated damage computation



STUDY B: UMBILICAL CABLE DAMAGE MONITORING (II)

Model	Curvature RMSE [m ⁻¹]		Pearson correlation	
	mean	Std	mean	std
Linear regression	0.00239	0.0044	0.726	0.342
Ridge	0.00098	0.0014	0.859	0.239
Lasso	0.00104	0.0015	0.864	0.219
Random forest	0.00107	0.0019	0.874	0.221
ANN	0.00138	0.0018	0.757	0.321

Model Results

- Regularized models (Ridge, Lasso) and RF - best performance
- Ridge - lowest RMSE
- Random Forest - highest Pearson correlation
- Linear Regression - underperformed, insufficient model flexibility
- ANNs likely suffered from limited training instances

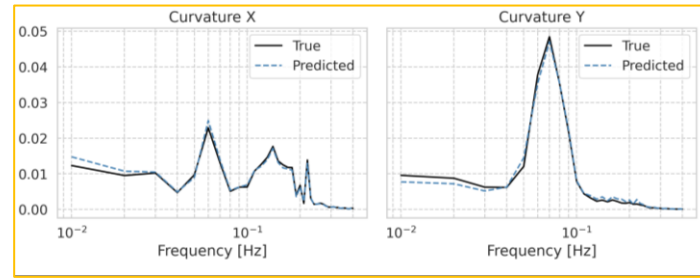
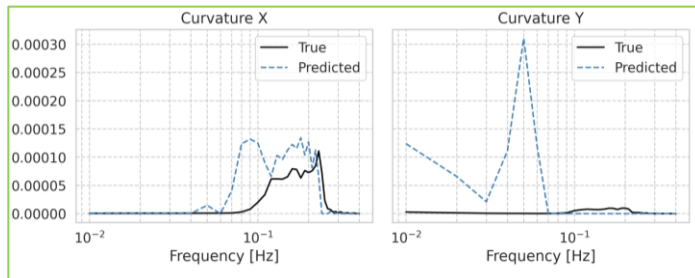
Model Performance Across Conditions

Hs = 0.25 m-Low energy

- Low absolute error
- High relative error
- Worse reconstruction

Hs = 6.75 m-High energy

- High absolute error
- Low relative error
- Better reconstruction



Model accuracy depends on sea state energy, but model ranking remains consistent

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Combining **HarshLab data**, **physics-based models** and **AI** enables more reliable **degradation assessment** and RUL estimation for **offshore renewable systems**.

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