

# POMDP-Based Risk Maintenance Planning for Offshore Wind Substructures

Pablo G. MORATO, Jannie S. NIELSEN, Philippe RIGO

pgmorato@uliege.be +32 4 366 48 57

## PARTNERS



## ABSTRACT

This work presents a novel methodology to identify the optimal maintenance strategy for an offshore wind structural component, providing a flexible and reliable support to decision-making and balancing inspection, repair and failure costs.

The methodology is tested for a tubular joint through a 60-states POMDP, obtaining the optimal maintenance policy in low computational time and in good agreement with common Risk-Based Inspection (RBI) methods.

## 1. INTRODUCTION

### Context:

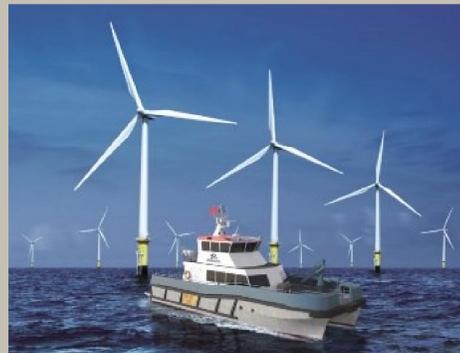
Wind farms farther from shore  
Complicated maintenance tasks

### Research Aim:

To identify the optimal maintenance strategy

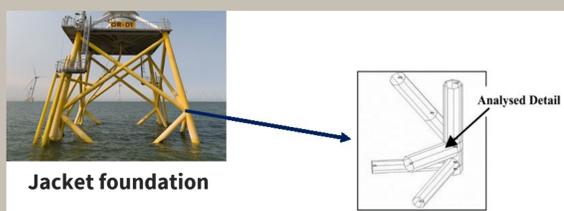
### Impact:

O&M cost ( $\approx 25\%$  LCOE)  
Lifetime Extension



Offshore wind crew transfer vessel

## 2. FATIGUE DETERIORATION MODEL



Jacket foundation

Structural joint

The calibration of the fracture mechanics (FM) model based on the SN-Miner's model provides a deterioration framework where inspections outcomes can be incorporated while keeping the model related to SN empirical data which is employed during the design stage.

### SN-Miner's model - Limit state:

$$g_{SN(t)} = \Delta - \frac{v_0 t}{\eta} \left[ \frac{q^{m_1}}{a_1} \Gamma \left( 1 + \frac{m_1}{h}; \left( \frac{S_1}{q} \right)^h \right) + \frac{q^{m_2}}{a_2} \gamma \left( 1 + \frac{m_2}{h}; \left( \frac{S_1}{q} \right)^h \right) \right]$$

### Fracture mechanics model - Limit state:

$$g_{FM(t)} = a_c - \left[ \left( 1 - \frac{m}{2} \right) C K^m \pi^{\frac{m}{2}} q^m \Gamma \left( 1 + \frac{m}{h} \right) \Delta n + a_{t-1} \left( 1 - \frac{m}{2} \right)^{\frac{2}{2-m}} \right];$$

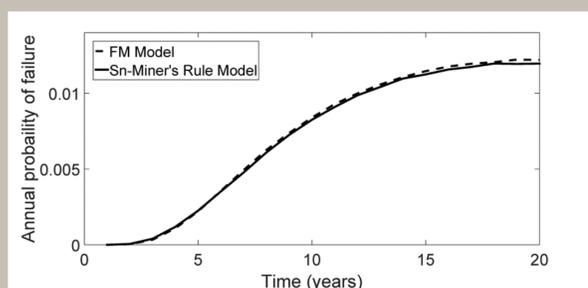
given  $a_0$

### SN-Miner's model - Variables

Variable	Distribution	Parameters
$m_1$	Deterministic	3
$m_2$	Deterministic	5
$\log_{10}(a_1)$	Normal	$\mu = 12.88$ ; $SD = 0.2$
$\log_{10}(a_2)$	Normal	* Fully correlated with $\log_{10}(a_1)$
$\Delta$	Lognormal	$\mu = 1$ ; $CoV = 0.3$
$q$	Weibull	$\mu = 13.0$ ; $CoV = 0.25$
$h$	Deterministic	0.8

### Fracture mechanics model - Variables

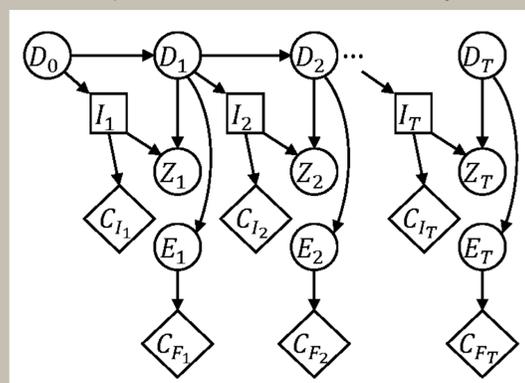
Variable	Distribution	Parameters
$a_0$	Exponential	$\mu = 0.057$
$\log(C)$	Deterministic	$\mu = -12.7$ ; $CoV = 0.19$
$q$ (*)	Normal	$\mu = 13.0$ ; $CoV = 0.25$
$h$	Deterministic	0.8
$v_0$	Deterministic	5045760
$K$	Deterministic	1
$m$	Deterministic	3
$a_c$	Deterministic	25



Calibration fracture mechanics - SN/Miner - reliability

## 3. RISK-BASED POMDP MODEL

The influence diagram below displays how the sequential decision problem is approached. The damage evolving over time is represented by the chance node  $D_t$  and it is possible to choose an inspection method in the node  $I_t$ .

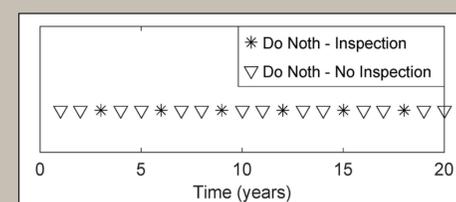


Sequential decision problem

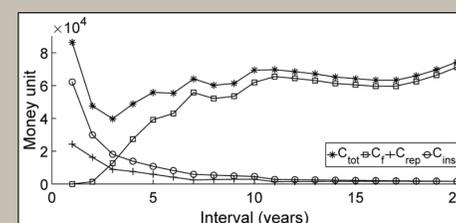
The chance node  $Z_t$  indicates the quality of the inspection method. Additionally, the node  $E_t$  tracks the failure probability. The utility nodes  $C_F$  and  $C_I$  assign a cost of failure and a cost of inspection, respectively. The chance node  $R_t$  represents the decision of whether to perform a repair or not.

## 4. RESULTS - POMDP POLICY

The optimal maintenance strategy for a tubular joint is identified by a 60-states "point-based" Partially Observable Markov Decision Process (POMDP). The obtained POMDP policy provides similar results as a common risk-based heuristic model.



POMDP Policy



Heuristic Policies

## 5. CONCLUSION

The 60 states infinite horizon POMDP has been solved providing the optimal maintenance policy for a tubular joint in only **0.32 seconds of computational time**.