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# ACTIVE TRUNCATION OF SLENDER MARINE STRUCTURES: INFLUENCE OF THE CONTROL SYSTEM ON FIDELITY

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## Data Sheet

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# Active truncation of slender marine structures: influence of the control system on fidelity

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## Abstract

Performing hydrodynamic model testing of ultra-deep water floating systems at a reasonable scale is challenging, due to the limited space available in existing laboratories and to the large spatial extent of the slender marine structures that connect the floater to the seabed. In this paper, we consider a method based on real-time hybrid model testing, namely the *active truncation* of the slender marine structures: while their upper part is modelled physically in an ocean basin, their lower part is simulated by an adequate numerical model. The control system connecting the two substructures inevitably introduces artefacts, such as noise, biases and time delays, whose probabilistic description is assumed to be known. We investigate specifically how these artefacts influence the fidelity of the active truncation setup, that is its capability to reproduce correctly the dynamic behaviour of the system under study. We propose a systematic numerical method to rank the artefacts according to their influence on the fidelity of the test. The method is demonstrated on the active truncation of a taut polyester mooring line.

## 1 Introduction

Floating structures used in the oil&gas, offshore wind or aquaculture industries require significant investments and must operate according to high safety and environmental standards. Therefore, the design of such structures is in general verified by means of hydrodynamic model testing prior to their construction. When performing such laboratory testing, the floating structure under study is constructed at reduced scale, and exposed to selected environmental conditions (wave, wind and current) that may be experienced during its design life. It is verified that the motions of the platform, the loads in the mooring and riser systems, or other quantities of interest (QoI) are acceptable under these conditions. The test campaign is in general also a final risk mitigation campaign, during which events not yet

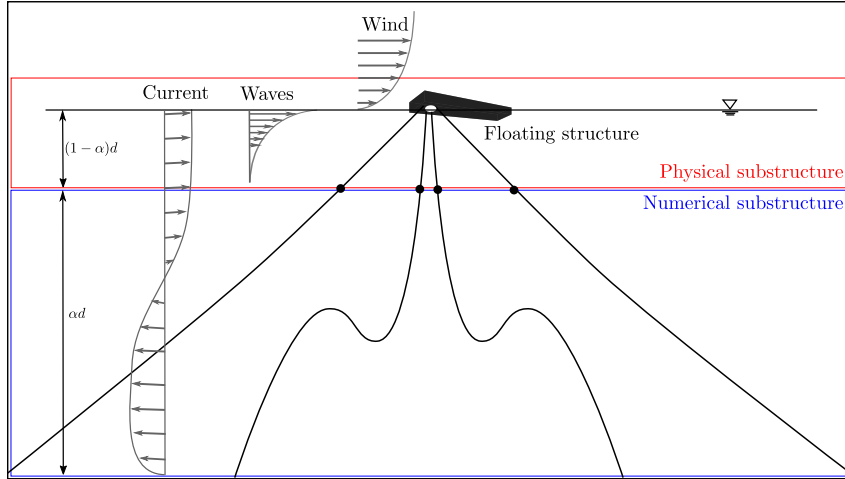


Figure 1: Model testing of an offshore structure with taut mooring and a flexible riser system in water depth  $d$ . Illustration of active truncation with truncation ratio  $\alpha = 0.8$ .

fully described by engineering numerical tools, such as green water and wave impact, could be detected and analyzed Pakozdi et al. (2017).

Floating structures are, however, installed at locations with increasingly large water depths. Oil exploitation takes nowadays place down to nearly 3000 m water depth Bowers (2016), and deep-sea mining of minerals is considered in water depths reaching 6000 m Sharma (2017). Modelling such systems with a reasonably large scale factor in existing hydrodynamic laboratories is challenging, due to both the vertical extent of the mooring system, but also due to its horizontal footprint, that ranges from two to four times the water depth Randolph and Gouvernec (2017). This challenge has been identified two decades ago, and has been addressed in details in Stansberg et al. (2002). The state-of-the-art approach, up to now, consists in performing *passive truncation* of the slender marine structures, as described briefly in the following. In a first stage, a truncated version of the mooring/riser system is designed such that it is *statically equivalent* to the full-depth system, and fits in the ocean basin Fylling and Stansberg (2005). It should be emphasized that the *dynamic* properties of the truncated system, such as the level of drag-induced damping of the horizontal motions of the floater, are generally not representative of the full-depth system, except possibly on a narrow range of sea-states. Model testing is then performed using the truncated system, and the experimental results are used to calibrate a numerical hydrodynamic model *of the floater* connected to the truncated system. The truncated system is finally replaced by the full-depth one in the *numerical* analysis, and the QoI, such as extreme motions and mooring line tensions, are evaluated numerically. In spite of recent improvements in the truncation procedures, which have been reviewed for example in Sauder et al. (2017); Cao and Tahchiev (2013), passive truncation still requires to calibrate a numerical model of the floater, which is time consuming and induces additional uncertainties. Furthermore, since the truncated

system used in the model tests is only statically equivalent to the full-depth system, it can be argued that some highly nonlinear effects driven by the floater’s dynamics (such as the occurrence of negative air gap or green water on deck) could remain undetected.

In the present paper, we consider an alternative solution denoted *active truncation*. It is based on the ReaTHM<sup>®</sup> testing<sup>1</sup> paradigm, already applied to solve issues related to model testing of floating wind turbines Sauder et al. (2016), and with applications beyond the field of marine technology Edrington et al. (2015); McCrum and Williams (2016). When performing active truncation, the floating structure and the upper part of the slender structure system are modelled *physically* in the ocean basin, while its lower part, which does not fit in the basin, is *simulated* on a computer. This is illustrated in Fig. 1. At the *truncation point*, the numerical and the physical *substructures* interact through a *control system*, including sensors and actuators. Therefore, active truncation intrinsically represents the full-scale system, and allows to obtain the QoI directly after the test, without the need for numerical extrapolation. Note that a strict pre-requisite to perform active truncation is the *validity of the numerical model describing the truncated portion of the slender marine structure*. In most cases, state-of-the-art programs based on the nonlinear Finite Element (FE) method can describe the low-frequency and wave-frequency dynamics of slender marine structures in a satisfactory manner, as for example illustrated in Figure 2 in Aksnes et al. (2015). However, some phenomena, such as complex soil-structure interaction or Vortex-Induced Vibrations (VIV) can still not be simulated with a sufficiently high level of confidence, at least not in real-time. This means that, as of today, if these phenomena are very subject or play a significant role in the empirical study, ReaTHM testing can not be applied.

The uncertainties that affect purely empirical and numerical approaches have been extensively studied in the past Qiu et al. (2014); ASME (ed.) (2016). However, when performing active truncation (and ReaTHM testing in general), a new source of uncertainty should be considered, namely the one originating from an imperfect *coupling* between the substructures. Indeed, various types of *artefacts*, such as noise, biases and time delays, are inevitably introduced by the presence of the control system Vilsen et al. (2017). Such artefacts, could jeopardize the *fidelity* of the setup, in the sense that they could make the system’s dynamical properties deviate significantly from those of the real system under study. In the worst case, this could happen without the operator of the test, or the final user of the empirical data, being aware of it. In this paper, we will neglect the uncertainties related to the physical and numerical substructures, to isolate and focus on those related to the control system.

This paper proposes a quantitative definition of fidelity, and presents a method to evaluate it for an active truncation setup. We then show how to systematically *identify the control system-induced artefacts that jeopardize the most the fidelity* (sensitivity study). This lat-

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<sup>1</sup>ReaTHM<sup>®</sup> testing stands for "Real-Time Hybrid Model testing", and is a registered trademark of SINTEF Ocean AS.

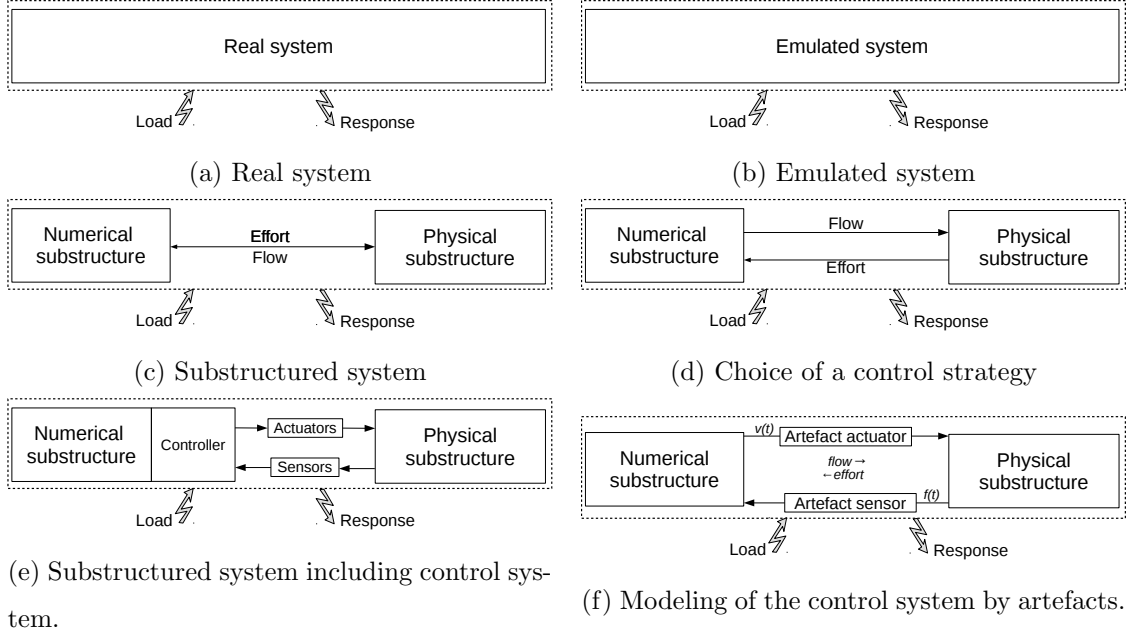


Figure 2: The various steps, and associated terminology, in the design and analysis of real-time hybrid model testing in general, and active truncation in particular.

ter aspect is believed to be a significant scientific contribution, in addition to be of great operational relevance when such testing methods are to be applied in practice.

The paper is organized as follows. In Section 2, a general method for the analysis of fidelity is outlined, and we show how it can be applied to the active truncation of slender marine structures. This method requires the capability of simulating an active truncated setup, including artefacts, which is the object of Section 3. In Section 4, the method is demonstrated on the truncation of a taut polyester mooring line, which is a widely used component for the positioning of offshore structures in deep water.

## 2 Fidelity analysis and its application to active truncation

In this Section, we first introduce some concepts and terminology which will be used throughout the paper. We then define a quantitative measure of fidelity, and outline a general method to evaluate it and study its sensitivity to artefacts. We finally show how it can be applied to address the active truncation problem.

### 2.1 Background and terminology

The *real system* (Fig. 2a) is the subject of the study, whose performance under given load conditions should be documented. It is for example the marine system (floater, mooring

and riser) represented in Figure 1. For analysis purposes, it is assumed that the real system can be fully represented by an *emulated system* (Fig. 2b). The emulated system consists of a numerical model capable of simulating the behaviour of the real system in a wide range of operational conditions, including extreme environmental conditions. For slender marine structures, the requirements and nature of this model strongly depends on the considered problem. Indeed, when VIV are neglected, top tensioned risers or taut polyester mooring lines in deep water can be satisfactorily simulated with linear time-domain (FE) methods, based on bar elements, and including nonlinear drag loads Rustad et al. (2008). Other types of structures, such as flexible risers or steel catenary risers, require the modelling of geometric nonlinearities, bending stiffness, and possibly nonlinear material properties and soil-structure interaction Fergestad and Løtveit (2015).

Performing active truncation consists in splitting the slender marine structure into two *substructures* located on either side of a *truncation point* (Fig. 1). The *truncation ratio*  $\alpha$  is defined as the ratio between the height of the water column occupied by the numerical substructure, and the total water depth  $d$ . At the truncation point, kinematic compatibility (equality of translational and rotational velocities) and dynamic equilibrium between the two substructures must be ensured at each instant. In more generic terms, the compatibility of *flow* and *effort* should be ensured at the interface between the substructure (Fig. 2c). To realize this in practice, a *control strategy* is chosen. As depicted in Fig. 2d, it can for instance be decided to *measure* the effort from the physical substructure (and prescribe it to the numerical substructure), and *prescribe* the flow (evaluated from the numerical substructure) to the physical substructure. In this way, the two substructures interact in real-time through a *control system* that includes sensors, actuators, as well as related software components such as force controllers and observers Vilsen et al. (2017) (Fig. 2e).

In reality, however, the control system inevitably introduces *artefacts*, such as measurement noise, time-delays due to communication, or imperfect actuation due to the actuators' own dynamics (Fig. 2f). The effects of some selected artefacts on a reference signal are illustrated in Fig. 3. Few authors have investigated the influence of such artefacts on the dynamics of substructured slender structures. The effect of interface time delays on substructured cables and beams have been studied analytically by Terkovic et al. (2016) and Zhang and Stepan (2016), respectively. However, in both cases, the authors focus on the *stability* of the substructured system only. While stability is indeed a necessary condition for the execution of active truncation, it does not guarantee that the active truncation setup represents the real system in a satisfactory way. In a recent study, Drazin et al. (2015) compared the displacement field of a substructured beam to the one of the original beam (the emulated system) by using an  $L^2$ -based error measure. In this work, the beams were described by Bernoulli-Euler equations and subjected to harmonic loading. The artefacts introduced at the interface were constant amplitude and phase mismatches, which modelled

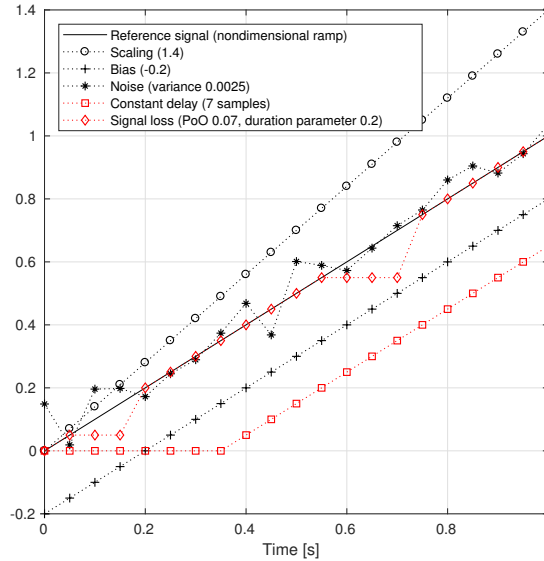


Figure 3: Various types of artefacts affecting a non-dimensional ramp signal, with their describing parameter in parentheses.

imperfect actuators.

Studying fidelity of the active truncation problem with similar analytical approaches is challenging when marine structures are involved. The first and main reason is that such structures must in most cases be described by *purely numerical* methods, such as the nonlinear FE method, which are difficult to exploit in analytical derivations. Analytical formulations could admittedly be obtained by making strong assumptions on the behaviour of the structure, but this would lead to an emulated system that does not necessarily reflect the *real* system anymore, and would make the resulting analysis questionable. The second reason is related to the fact that a control system introduces not only one, but *several* types of artefacts *at the same time*. Suitable frameworks exist for studying the individual effect of each of these artefacts: stochastic differential equations, delayed differential equations, networked control systems theories allow for example to study the effect of noise, delays, and jitter, respectively. However, *combining* these frameworks leads to formulations that are intractable in practice. Also, making simplifying assumptions in this regard, by for example considering only one selected type of artefact, is questionable, since it is unclear *a priori* which artefact jeopardizes the fidelity, and which one can be neglected, if any. In the following subsection, we will outline a method to address these two issues.

## 2.2 Fidelity analysis method

**Proposed definition of fidelity** The fidelity  $\varphi$  is calculated by comparing selected QoI, evaluated on the one hand from the emulated system (Fig. 2b), and on the other hand from the substructured system *including artefacts* (Fig 2f), when these two systems are subjected to the *same* external excitation. The comparison function is chosen so that

the value of  $\varphi$  tends to infinity when the QoI for the two systems are identical, and takes low values when the artefacts make the substructured and the emulated systems differ significantly. The selection of the external excitation, QoI, and comparison function is problem dependent, and we will propose a definition applicable to our active truncation problem in the next subsection.

**Modelling of artefacts** As indicated in the legend of Fig. 3, each artefact is characterized by one or several parameters. For instance, white noise is described by its variance, and signal loss is described by both its probability of occurrence and its duration parameter. By gathering these parameters, the *heterogeneous set* of artefacts affecting the substructured system can be parametrized by a single  $M$ -components vector  $\theta$ . In practice, the amount of noise or time delays present in an active truncation setup cannot be perfectly known until the setup has been realized. It is therefore considered that  $\theta$  is the realization of a random vector  $\Theta$ , with joint probability density function  $f_{\Theta}(\theta)$ . In the scope of this paper, the components of  $\Theta$  are assumed to be statistically independent.

For a given realization  $\theta$  of the artefacts' parameter, the fidelity  $\varphi(\theta)$  can then be evaluated from *co-simulations* of the substructured system. The term co-simulation is used, since in the analysis, the physical and the numerical substructures are represented by *separate* numerical models, which are *coupled* at the truncation point in a dynamic simulation that includes the effect of the artefact. In Section 3, we will detail how this co-simulation is performed for slender marine structures.

**Polynomial chaos expansions** Due to the random nature of the artefacts' parameter  $\Theta$ , the fidelity  $\varphi(\Theta)$  will also be a random variable, whose variance is assumed to be finite. It can then be approximated by the following (truncated) polynomial chaos expansion (PCE) Xiu and Karniadakis (2003):

$$\hat{\varphi}(\Theta) := \sum_{\alpha \in \mathcal{A}} a_{\alpha} \psi_{\alpha}(\Theta) \quad (1)$$

where  $\mathcal{A}$  a finite subset of  $\mathbb{N}^M$ ,  $(a_{\alpha})_{\alpha \in \mathcal{A}}$  is a family of real numbers,  $(\psi_{\alpha})_{\alpha \in \mathbb{N}^M}$  a family of orthonormal polynomials with respect to the input variable  $\Theta$ , i.e.

$$\mathbb{E}[\psi_{\alpha}(\Theta)\psi_{\beta}(\Theta)] = \int_{\mathcal{D}} \psi_{\alpha}(\theta)\psi_{\beta}(\theta)f_{\Theta}(\theta)d\theta := \delta_{\alpha\beta} \quad (2)$$

where  $\delta_{\alpha\beta}$  is the Kronecker delta. Note that since the orthogonality condition in (2) depends on  $f_{\Theta}(\theta)$ , so does the chosen family of polynomials  $(\psi_{\alpha})$  in (1). We will demonstrate by an example in Section 4 how  $\hat{\varphi}$  is determined in practice.

**Uncertainty propagation** Thanks to the orthogonality property (2), estimates of  $\mathbb{E}[\varphi(\Theta)]$  and of  $\text{Var}[\varphi(\Theta)]$  can be obtained from the  $a_{\alpha}$  coefficients by:

$$\mathbb{E}[\hat{\varphi}(\Theta)] = a_0 \quad (3)$$

$$\text{Var}[\hat{\varphi}(\Theta)] = \sum_{\alpha \in \mathcal{A} \setminus \{0\}} a_{\alpha}^2 \quad (4)$$

These estimates can then be related to a minimum admissible fidelity  $\varphi_{adm}$ , defined by the experimentalist, or the final user of the test results. Note that the value of  $\varphi_{adm}$  will depend on the exact definition of the fidelity, and will differ depending on the purpose of the test. For instance, if the active truncation tests aim at the final verification of a concept, one will aim at a “high” fidelity, and therefore a high value of  $\varphi_{adm}$ . If, on the contrary, they are related to a preliminary feasibility study, lower values of  $\varphi_{adm}$  could be accepted.

**Sensitivity analysis** If  $E[\varphi(\Theta)]$  is deemed too low, or the uncertainty  $\text{Var}[\varphi(\Theta)]$  too large, the natural course of action is to determine which artefacts influence the most the variations of the fidelity. To do so, we will use the variance decomposition method (ANOVA), based on the Sobol’ decomposition Sobol (1993). Under the assumption of finite variance, which we assume to be fulfilled for our physical problem, the following decomposition exists and is unique:

$$\varphi(\theta) = \varphi_0 + \sum_{i=1}^M \varphi_i(\theta_i) + \sum_{1 \leq i < j \leq M} \varphi_{i,j}(\theta_i, \theta_j) + \dots + \varphi_{1,2,\dots,M}(\theta_1, \theta_2, \dots, \theta_M) \quad (5)$$

where  $\varphi_0$  is constant, and the integral of each summand over any of its independent variables is zero. In this setting,  $\text{Var}[\varphi(\Theta)] = \sum_{i=1}^M V_i + \sum_{1 \leq i < j \leq M} V_{i,j} + \dots + V_{1,2,\dots,M}$ , where each term corresponds to the variance of the corresponding term in (5). Normalizing the above decomposition by  $\text{Var}[\varphi(\Theta)]$ , the *Sobol’ indices* are defined, which satisfy

$$\sum_{i=1}^M S_i + \sum_{1 \leq i < j \leq M} S_{i,j} + \dots + S_{1,2,\dots,M} = 1 \quad (6)$$

The  $S_i$  are called *first-order* Sobol’ indices,  $S_{i,j}$  *second order* Sobol’ indices, etc... The *total* Sobol’ indices  $S_{T,i}$  are defined as the sum of all Sobol’ indices involving the  $i^{\text{th}}$  parameter  $\theta_i$ . By ranking the  $S_{T,i}$ , the  $\theta_i$  having the greatest impact on the variations of the fidelity can be identified. Also, by comparing each  $S_{T,i}$  to  $S_i$ , it is possible to evaluate whether parameter  $\theta_i$  influences  $\varphi$  alone (in the case  $S_i \approx S_{T,i}$ ), or jointly with other parameters of  $\Theta$ .

The evaluation of Sobol’ indices used to be computationally expensive, in the sense that numerous evaluations of  $\varphi(\theta)$ , and therefore numerous co-simulations of the substructured system, were required. However, it was recently shown in Sudret (2008) how Sobol’ indices could be computed analytically from the expansion (1). This result, associated with the significant advances on adaptive sparse PCEs Blatman and Sudret (2011), makes PCE a tool of choice for for uncertainty propagation and sensitivity analyses.

### 2.3 Fidelity indicators for the active truncation problem

Let us now show how this framework applies to our problem. We consider the active truncation of a mooring line connecting the floating structure to the sea bottom, as shown in

Figure 1. Without loss of generality, we assume that the problem is two dimensional, and we define a direct  $x$ - $z$  coordinate system, whose  $z$  axis is vertical and pointing upwards.

The fidelity will be evaluated by studying the response of the slender structure to a characteristic external load vector  $\tau(t)$ , with a duration  $T$ , in seconds. This load is meant to be representative, in terms of amplitude, frequency content and direction, of a severe load that can be encountered during the testing of a truncated mooring line. The *dynamic* part of this load represents wave loads transferred from the floater to the slender structure, and is therefore applied to the top of the slender marine structure. It has two components. The first *low-frequency* component acts mainly axially, has an amplitude of 1MN, and a frequency content sweeping  $[0, 0.02]$  Hz. It mimics the effect of second-order difference-frequency wave loads. The *wave-frequency* component has an amplitude of 250kN, and a frequency content sweeping  $[0, 0.2]$  Hz, and a direction with constant rate of change. This dynamic load comes in addition to the *static* top tension applied to the slender structure, and to the drag load associated to a shear current, whose velocity varies linearly throughout the water column for 0m/s at the seabed to 0.5m/s at the free surface. Time series of the top load can be seen in Fig. 5.

We will now focus on the definition of the fidelity indicator  $\varphi$  for the active truncation problem. In hydrodynamic model test campaigns, the focus is generally on the behaviour of the floater, and on extreme tensions in the slender marine structures, but not on their local deflection or curvature. The objective is therefore to make the interaction between the truncated slender marine structure, the (physical) floater and the (numerical) sea bottom reflect the corresponding interactions in a fully physical setup. In other words, *the exact behaviour of the slender structure throughout the water column is assumed of minor importance, as long as its interactions with the floater and the sea bottom are modelled properly.*

Based on this reasoning, two fidelity indicators are suggested. Let  $V_{x,top}$  and  $V_{z,top}$  be the components of the top velocity of the slender structure, and  $F_{x,bottom}$  and  $F_{z,bottom}$  the components of the force vector at its lower end. These values are calculated by co-simulation of the substructured system, that includes the artefacts parametrized by  $\theta$ . Let  $\bar{V}_{top}$  and  $\bar{F}_{bottom}$  be their ideal counterparts, obtained by simulation of the emulated system. Then, the first indicator

$$\varphi_1(\theta) = -\frac{1}{2} \log_{10} \left( \frac{\int_0^T (V_{x,top}(t|\theta) - \bar{V}_{x,top}(t))^2 dt}{\int_0^T \bar{V}_{x,top}(t)^2 dt} + \frac{\int_0^T (V_{z,top}(t|\theta) - \bar{V}_{z,top}(t))^2 dt}{\int_0^T \bar{V}_{z,top}(t)^2 dt} \right) \quad (7)$$

quantifies how well the top end of the structure responds to the prescribed external load  $\tau$ , and thus how well the substructured system manages to replicate the mechanical impedance of the slender structure.  $\varphi_1$  is therefore important when motions of the floater are investigated. The second indicator

$$\varphi_2(\theta) = -\frac{1}{2} \log_{10} \left( \frac{\int_0^T (F_{x,bottom}(t|\theta) - \bar{F}_{x,bottom}(t))^2 dt}{\int_0^T \bar{F}_{x,bottom}(t)^2 dt} + \frac{\int_0^T (F_{z,bottom}(t|\theta) - \bar{F}_{z,bottom}(t))^2 dt}{\int_0^T \bar{F}_{z,bottom}(t)^2 dt} \right) \quad (8)$$

Table 1: Properties of the polyester mooring line used in the case studies.

Parameter	Unit	Value
Length	m	1934
Diameter	mm	264
Mass per unit length	kg/m	44.7
Young modulus	GPa	8.513
Submerged weight per unit length	N/m	93.2
Rayleigh damping coefficient $\alpha_2$	s	$4.77 \cdot 10^{-2}$
Top tension module	kN	2500
Top tension angle	$^\circ$	50
Normal added mass coefficient	-	1.0
Tangential added mass coefficient	-	0.0
Normal drag coefficient	-	1.6
Tangential drag coefficient	-	0.0

quantifies how well the external load is transferred to the sea bottom, and is then more relevant when the focus is on loads on e.g. anchors or blow-out preventers. If both aspects are important,  $\varphi_1$  and  $\varphi_2$  could easily be combined into a single indicator.

To summarize, in this Section, we have (1) suggested two possible expressions of the fidelity  $\varphi$  for the active truncation problem. (2) We discussed how  $\varphi$  could be jeopardized by heterogeneous and random artefacts, described by a random vector  $\Theta$ . (3) We showed how  $E[\varphi(\Theta)]$  and  $\text{Var}[\varphi(\Theta)]$  could be evaluated (uncertainty propagation) from the PCE of  $\varphi$ , and (4) we introduced the Sobol' indices characterizing the sensitivity of  $\varphi$  to the various components of  $\Theta$ . This analysis method will be demonstrated by a practical case study in Section 4. This case study requires the ability to co-simulate an active truncation setup including artefacts. This will be the object of the next section.

### 3 Co-simulation of slender marine structures including artefacts

The first part of this section describes a method to co-simulate the system presented in Fig. 2f, when the substructures are slender marine structure. As an example, the taut polyester mooring line, whose properties are given in Table 1, will be substructured, and a co-simulation will be performed, corresponding to a deterministic value of  $\theta$ , to put in evidence the effect of selected artefacts on the dynamics of the system.

### 3.1 The fixedFreeCableSegment model

The FE method is used to simulate the slender marine structure. The analysis is two-dimensional, and the structure is represented by a bar element model (2 degrees of freedom per node) as shown in Fig. 4a. The boundary condition of the structure is *fixed-free* which means that the velocity of lower end of the structure, and the force on the upper end, are prescribed. Inertia, added-mass, drag and effective weight loads are included in a similar way as in Rustad et al. (2008). The stiffness matrix has both an elastic and a geometric component. Since the geometric component strongly depends on the configuration of the structure, the static equilibrium is found by Newton-Raphson iterations.

The dynamic analysis is linear, in the sense that it uses the mass matrix  $M$  and the stiffness matrix  $K$  determined by the static analysis, throughout the time domain simulation. Nonlinearities due to drag loads are modelled exactly. These modelling choices are adequate to simulate structures with minimal changes of configuration, such as top tensioned risers or taut mooring lines in deep water, for which lateral deflections are about two orders of magnitude smaller than the structure's length. The structural damping matrix is of the form  $C = \alpha_1 M + \alpha_2 K$  (Rayleigh damping) where  $\alpha_1$  is chosen to be null. In that case, the damping ratio associated to a vibration mode with circular frequency  $\omega_i$  is  $\lambda_i = \omega_i \alpha_2 / 2$ . The model is implemented as a MATLAB<sup>®</sup> class named `fixedFreeCableSegment`. The verification of this class is presented in A. It is also shown how the eigenmodes of the taut polyester mooring differs from those of a string, due to the combined effects of elasticity, varying tension, and oblique configuration, which would be inconvenient to represent in a purely analytical model.

In the following, we will show how an active truncation setup can be modelled by coupling two such `fixedFreeCableSegment` objects.

### 3.2 Co-simulation without artefacts

The active truncation setup is represented in Fig. 4b. The water depth is  $d = 1200$  m, and the truncation ratio is  $\alpha = 0.8$ . The physical substructure (in red) and the numerical substructure (in blue), denoted  $\mathbf{p}$  and  $\mathbf{n}$ , respectively, are *each* modelled by a `fixedFreeCableStructure` object. The top velocity  $V_{top}$  in (7) will hence be evaluated from  $\mathbf{p}$ , and the bottom force  $F_{bottom}$  in (8) from  $\mathbf{n}$ . Focusing now on the truncation point, the selected boundary conditions in `fixedFreeCableSegment` are such that the bottom velocity of  $\mathbf{p}$  and the top force acting on  $\mathbf{n}$  can be prescribed. Their dual values, that is the bottom force on  $\mathbf{p}$ , and the top velocity of  $\mathbf{n}$ , can be evaluated by time integration. The dynamic equilibrium and kinematic compatibility at the truncation point is satisfied by the iterative procedure described in Algorithm 1.

There are three important parameters in this algorithm. The *synchronization time step*  $\delta t$  is the duration between two time instants at which equilibrium and compatibility at the

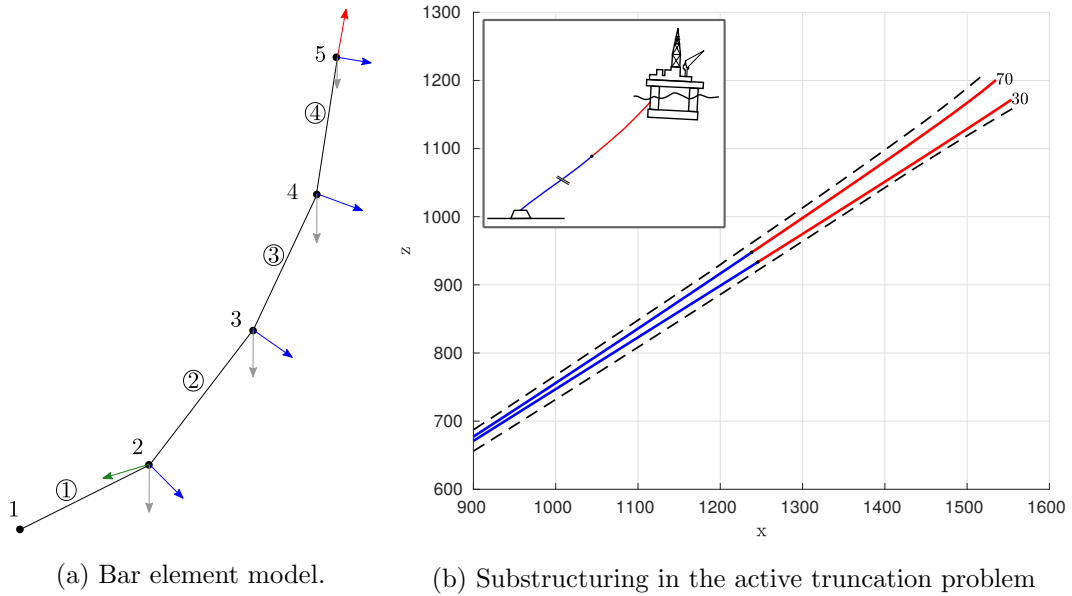


Figure 4: Subfigure (a): bar element model as implemented in the `fixedFreeCableSegment` class. Element numbers are circled. Nodal forces are represented by arrows: effective weight force (grey), current forces (blue), prescribed external top force (red) and prescribed bottom displacement (green). Subfigure (b): overview of active truncation problem on the top left, with the physical substructure in red and the numerical substructure in blue. Main plot: snapshots of the upper part of the polyester line at  $t = 30s$  and  $t = 70s$ , when subjected to the characteristic excitation  $\tau$ . The dashed lines correspond to the envelope of the line's displacement during the analysis.

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**Algorithm 1** Co-simulation of two coupled `fixedFreeCableSegment` objects, denoted `n` and `p`.

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```
1: for  $t \in \{0, \delta t, \dots, T - \delta t\}$  do
2:  $v \leftarrow$  top velocity of n at time instant  $t$ 
3:  $v\_next \leftarrow \infty$ 
4:  $f \leftarrow$  bottom force of p at time instant  $t$ 
5:  $f\_next \leftarrow \infty$ 
6:   while true do
7: Perform time-integration of p from  $t$  to  $t + \delta t$  with varying external excitation and bottom velocity varying linearly to  $v$ 
8:  $f\_next \leftarrow$  bottom force of p at  $t + \delta t$ 
9: Perform time-integration of n from  $t$  to  $t + \delta t$  with varying external excitation and top force varying linearly to  $f\_next$ 
10:  $v\_next \leftarrow$  top velocity of n at  $t + \delta t$ 
11:   if  $\|v\_next - v\|_\infty > \epsilon_v$  OR  $\|f\_next - f\|_\infty > \epsilon_f$  then:  $v \leftarrow v\_next$  ;  $f \leftarrow f\_next$ 
12:   else: Jump to next synchronization time step
13:   end if
14: end while
15: end for
```

---

truncation point are enforced. In practice,  $\delta t$  will be chosen equal to the minimum loop time of the control system orchestrating the active truncation. During the iterations (lines 6-14),  $\epsilon_f$  and  $\epsilon_v$  are force and velocity tolerances, below which dynamic equilibrium and kinematic compatibility at the truncation point are assumed to be achieved, respectively. These parameters influence both the results and the computational time of a co-simulation, in the same way as the number of elements  $n_{el}$ , so their value must be chosen carefully. To this end, a convergence study is performed and reported in B. The selected values are  $n_{el}=80$  elements,  $\delta t=10$  ms,  $\epsilon_v=10^{-6}$  m/s and  $\epsilon_f=0.1$  N.

We have outlined how a co-simulation could be performed that satisfies equilibrium and compatibility criteria at the truncation point. Performing the fidelity analysis described in Section 2 requires now artefacts to be introduced in this coupling, which will be the object of the next subsection.

### 3.3 Co-simulation including artefacts

An `artefact` class was developed, which allows simulating calibration errors (multiplicative errors), bias (additive errors), white noise, delay, zero-order hold and signal loss. The class has a `signalIn` method to get an input, a `signalOut` method to retrieve an output, and in the particular case when no artefact should affect the signal, it works simply as a First-In-First-Out (FIFO) queue. When artefacts are present, the input is modified before being returned. As an example, in Fig. 3, successive calls to `signalOut` were made on `artefact` objects with different properties, which received identical samples of the reference signal via

the `signalIn` method.

As shown in Fig. 2f, two `artefacts` objects are needed, one acting on the effort (here, force) obtained from the experimental substructure, and the other one acting on the flow (here, velocity) obtained from the numerical substructure. Because they act on signals which are obtained from sensors, or used as reference to actuators, they will be denoted `aS` and `aA`, respectively. In this setting, performing a co-simulation that includes the effect of these artefacts requires only minor modifications to Alg. 1. (1) At line 8, `f_next` should be input to `aS.signalIn`, and the output of `aS.signalOut` should be used instead of `f_next` in line 9. (2) Similarly, `v_next` should be passed through `aA` after line 10 before being used. (3) At line 11, the convergence criterion should be evaluated on the values *affected by the artefacts*.

### 3.4 Example: effect of signal loss in active truncation

We will now illustrate the capabilities of this algorithm, and of the `artefact` and `fixedFreeCableSegment` classes with an example. We will consider a co-simulation in which *signal loss* affects both the measured force and the applied velocity. Signal loss may for example be due to sensor and communication issues, or to unfinished calculations in the numerical substructure Vilsen et al. (2017). It is parametrized by a probability of occurrence  $\zeta_1 \in [0, 1)$  and a characteristic duration parameter  $\zeta_2 > 0$ . The duration  $D$  of the signal loss (during which the signal is “frozen” to the last received value) is modelled as a random variable distributed as  $f_D(d) = e^{-\zeta_2 d} / \zeta_2$ . With this model, longer signal loss durations are expected for smaller values of  $\zeta_2$ . In the present case,  $\zeta_1$  is set to 1%, and  $\zeta_2$  to 0.1. So in this case, the artefacts can be parametrized by  $M = 4$  components, and  $\theta = (1\%, 0.1, 1\%, 0.1)^\top$ .

The mooring line is subjected to the characteristic excitation  $\tau$  described in Section 2. The results of the co-simulation are presented in Fig. 5. The dynamic excitation at the top node (first row) is identical for the emulated structure (black) and the physical substructure (red). This disturbance travels along the physical substructure (`p`), and reaches the truncation point where a force is measured (second row, red line). This force is possibly subjected to signal loss (`aS`) before being transferred, as a top force, to the numerical substructure (second row, blue line). The numerical substructure (`n`) responds to this top force (third row, blue line), and this response, which may also be affected by some signal loss (`aA`), is used to command the bottom velocity of the physical substructure (third row, red line). The effect of these signal loss on the QoI, which are the top velocity and the bottom force, are shown in the fourth and fifth rows, respectively. The fidelity is evaluated by comparing these signals to the ones obtained with the emulated system (black lines). By applying (7) and (8), it is found that  $\varphi_1 = 1.30$  and  $\varphi_2 = 1.99$ .

The right column in Fig. 5 shows a selected time window during which signal losses happen on the measured force and on the applied velocity. On the second row, we can for instance observe that the signal of the force sensor freezes for about half a second shortly

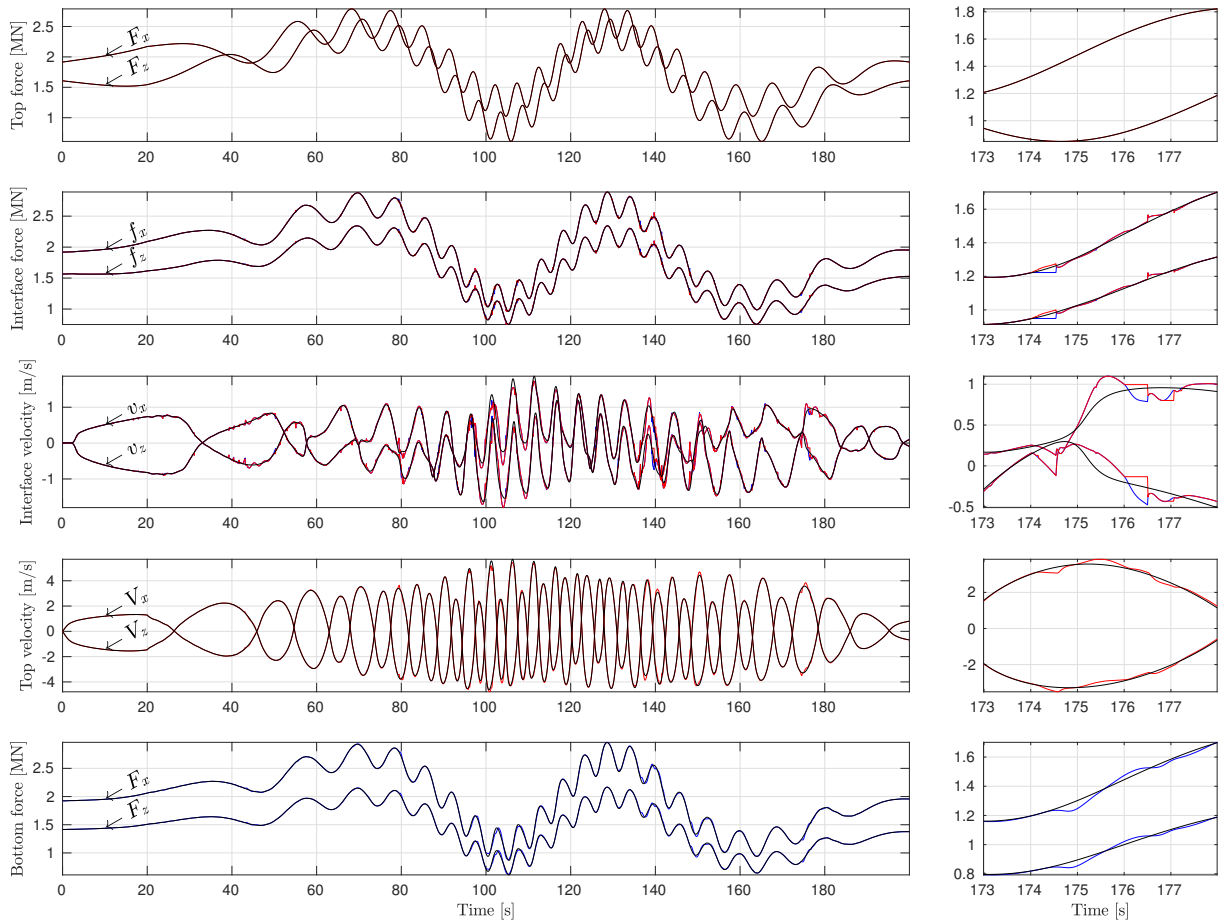


Figure 5: Active truncation setup subjected to the characteristic load  $\tau$ . The red curves are obtained from the physical substructure, and the blue curves from the numerical substructure. The black curves represent the emulated system. Signal loss occurs at the force sensors level (as visible on the second row) and at the velocity actuation level (third row). The right column is a zoom on the time series at a location of interest.

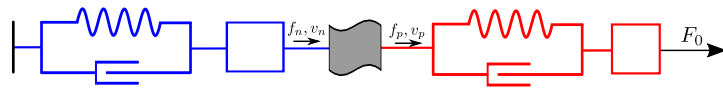


Figure 6: Stylized version of the active truncation problem, used to illustrate the effect of signal loss on the response of the substructures. The red mass-spring-damper (MSD) system represents the physical substructure, and the blue MSD system represents the numerical substructure. The flag-shaped box represents the truncation point, where signal loss occurs.  $F_0$  represents the top excitation.  $f$  and  $v$  represent the force and velocity at the truncation point, respectively:  $f_n$  and  $v_n$  are seen from the numerical substructure, and  $f_p$  and  $v_p$  from the physical substructure.

after  $t = 174$  s, since the red and blue lines differ from each other. On the third row, it can be seen that the velocity command signal freezes twice, first at  $t = 176$  s for half a second, then for about 300ms.

An important remark is that the substructure *from which the signal comes* has no direct information of the occurrence of a signal loss, but is anyway affected indirectly by the feedback it receives from the other substructure. Let us illustrate this by constructing a stylized version of our setup, represented in Fig. 6. We assume that signal loss occurs *on the force measurement only*, and that the velocity actuation is perfect, that is  $v_p = v_n$  at all times, while  $f_n \neq f_p$  when signal loss occurs. Starting from static equilibrium, when  $F_0$  increases, all other variables  $f_p$ ,  $f_n$ ,  $v_p$  and  $v_n$  will increase. If signal loss occurs in the force measurement,  $f_n$  keeps a constant value (instead of increasing), and  $v_n$  will eventually decrease due to the stiffness and damping of the numerical substructure. Since  $v_n$  directly steers the actuator command,  $v_p$  will decrease immediately, causing the stretching of the physical substructure, and an increase in  $f_p$ . When the signal on the force sensors is recovered,  $f_n$  will jump to the (larger)  $f_p$  value, causing a sudden increase of  $v_n$ , and thus  $v_p$ . This simplified example describes well the mechanism causing the significant decrease and increase of the velocity of the truncation point (on the physical substructure side) observed in Fig. 5 for  $t \in [174, 176]$  s. This perturbation propagates according to the nonlinear dynamics of the slender structure, to the top and bottom ends of the line, and is clearly observed both on the top velocity and bottom force time series. It will thus affect both  $\varphi_1$  and  $\varphi_2$ .

We have, in this Section, shown how the active truncation problem could be modelled, and with this last example, illustrated qualitatively the - possibly complex - interaction mechanisms resulting from e.g. signal losses at the truncation point. We will now extend the analysis to a larger set of *random* and *heterogeneous* artefacts, representing a more realistic case.

## 4 Fidelity analysis for a truncated taut mooring line

We will in this section show how the method presented in Section 2 can be applied to study the active truncation problem when multiple, heterogeneous and random artefacts are present. We consider the same polyester mooring line as in the previous example (see Table 1), installed in a water depth of  $d = 1200$  m. Active truncation is performed with  $\alpha = 0.8$ . We assume that the model tests are performed at a scale  $\lambda = 1/60$ . This means that the depth of the ocean basin laboratory, where sensors and actuators are installed, is  $(1 - \alpha)d\lambda = 4$  m. The two force components  $f_x$  and  $f_z$  at the truncation point are measured by two independent force sensors, and an actuator prescribes the velocity  $(v_x, v_z)$  of the truncation point. The mooring line is subjected to the characteristic load introduced in Section 2, and the fidelity indicators based on top velocity ( $\varphi_1$ ) and bottom force ( $\varphi_2$ ), defined in (7) and (8) are considered.

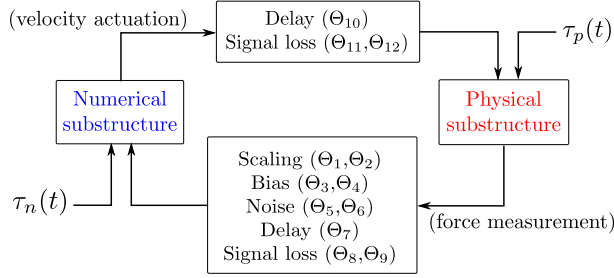


Figure 7: Structure of the system, modelled artefacts, and their describing parameters.  $\tau_n$  represents the current loads acting on the numerical substructure, and  $\tau_p$  represents the current loads and varying wave loads acting on the physical substructure.

As shown in Figure 7 and Table 2, ten individual artefacts, described by  $M = 12$  parameters, are assumed to affect the setup. The choice of including these artefacts, and neglecting others, is based on insight gained from the experimental work reported in Vilsen et al. (2017), but note that the core method would apply also if other artefacts were selected. Each component of the force measurement is assumed to be contaminated by calibration error, bias, and noise. In the acquisition process, the force signals can be delayed, or lost, before entering the numerical substructure. Signal loss at the output of the numerical substructure models the fact that the calculations in the numerical substructure may not complete on time. An additional delay on the actuation side models computation and communication processes. The probabilistic description of these artefacts is summarized in the last column of Table 2. Since only estimates of upper bounds, lower bounds, mean values, or standard deviations of the  $\theta_i$  parameters were available, the maximum entropy principle Jaynes (1957) was used to define  $f_{\Theta}(\theta)$ , which could be improved by dedicated surveys.

#### 4.1 LHS sampling and uncertainty propagation

As outlined in Section 2, the first objective is to estimate  $E[\varphi(\Theta)]$ , that is the expected fidelity for the active truncation setup, when it is affected by the set of artefacts described in Table 2.  $\text{Var}[\varphi(\Theta)]$  is also estimated, indicating how much the fidelity may vary due to the uncertainties on  $\Theta$ . As explained in Section 2, this is done by establishing a PCE surrogate model of  $\varphi$  (in the following,  $\varphi$  may designate either  $\varphi_1$  or  $\varphi_2$ ), denoted  $\hat{\varphi}$ , whose structure allows to evaluate efficiently  $E[\varphi(\Theta)]$  and  $\text{Var}[\varphi(\Theta)]$ . Such a surrogate model is a function of the twelve-dimensional variable  $\theta$ , and must mimic the behaviour of  $\varphi$  over its whole domain of definition. To establish  $\hat{\varphi}$ ,  $\varphi(\theta)$  must therefore be evaluated for a *space-filling* set of samples of  $\Theta$  denoted  $\mathcal{E}$ . This set is generated with the Latin Hypercube Sampling method (LHS), and  $\varphi$  is evaluated by co-simulation, as explained in Section 3, for each sample in  $\mathcal{E}$ . In Figure 8, the markers show 208 points generated by LHS in the twelve-dimensional space, with the associated value of  $\varphi_1$  plotted against each component of  $\theta$ . Note that these

Table 2: Description of the artefacts affecting the setup, including their probabilistic description.  $\mathcal{U}(a, b)$  refers to the uniform distribution with support  $[a, b]$ .  $\mathcal{N}(\mu, \sigma)$  refers to the normal distribution with mean  $\mu$  and variance  $\sigma^2$ . Here  $\lambda=1/60$  and  $\delta t=10$  ms.

Type of artefact	Affected signal	Describing parameter(s)	Unit	Probabilistic description
Calibration error	$f_x$	$\Theta_1$ (scaling factor)	-	$\mathcal{N}(1, 0.015)$
Calibration error	$f_z$	$\Theta_2$ (scaling factor)	-	$\mathcal{N}(1, 0.015)$
Bias	$f_x$	$\Theta_3$ (bias value)	N	$\mathcal{N}(0, 0.05\lambda^{-3})$
Bias	$f_z$	$\Theta_4$ (bias value)	N	$\mathcal{N}(0, 0.05\lambda^{-3})$
Noise	$f_x$	$\Theta_5$ (noise variance)	$\text{N}^2$	$\mathcal{U}((0.025\lambda^{-3})^2, (0.05\lambda^{-3})^2)$
Noise	$f_z$	$\Theta_6$ (noise variance)	$\text{N}^2$	$\mathcal{U}((0.025\lambda^{-3})^2, (0.05\lambda^{-3})^2)$
Delay	$f_x, f_z$	$\Theta_7$ (duration)	s	$\mathcal{U}(0, 5\delta t)$
Signal loss	$f_x, f_z$	$\Theta_8$ (probability of occurrence) $\Theta_9$ (duration parameter)	- $\text{s}^{-1}$	$\mathcal{U}(1\%, 10\%)$ $\mathcal{U}(0.1, 0.5)$
Delay	$v_x, v_z$	$\Theta_{10}$ (duration)	s	$\mathcal{U}(0, 5\delta t)$
Signal loss	$v_x, v_z$	$\Theta_{11}$ (probability of occurrence) $\Theta_{12}$ (duration parameter)	- $\text{s}^{-1}$	$\mathcal{U}(1\%, 10\%)$ $\mathcal{U}(0.1, 0.5)$

co-simulations are independent from each other and can therefore be performed in parallel. Also, nested LHS can be used Blatman and Sudret (2010), to sequentially add samples to  $\mathcal{E}$ , while ensuring that the updated set  $\mathcal{E}$  contain samples still distributed according to  $f_{\Theta}(\theta)$ . The distribution of  $\varphi$  can be estimated from  $\mathcal{E}$  (see Figure 9), and in Figure 10, the realization of  $\Theta$  leading to the median value of  $\varphi_1$  is shown for illustration.

Based on this initial set  $\mathcal{E}$  and on the associated values of  $\varphi$ , the PCE model  $\hat{\varphi}$  in (1) is established by using a degree-adaptive sparse PCE, based on least-angle regression (LARS, Blatman and Sudret (2011)), implemented in the UQLab software Marelli and Sudret (2014, 2017). These two references may be consulted by the interested reader for more details on the theoretical and practical aspects of PCE identification. The values of  $E[\varphi(\Theta)]$  and  $\text{Var}[\varphi(\Theta)]$  can then be evaluated from (3) and (4):

$$E[\hat{\varphi}_1(\Theta)] = 1.32 \text{ and } \text{Var}[\hat{\varphi}_1(\Theta)] = 0.13^2$$

$$E[\hat{\varphi}_2(\Theta)] = 1.77 \text{ and } \text{Var}[\hat{\varphi}_2(\Theta)] = 0.17^2$$

This means that the active truncation scenario selected in Figure 10, where  $\varphi_1=1.33$  and  $\varphi_2=1.62$ , corresponds to an *average fidelity* for the top velocity of the slender structure (when compared to  $E[\varphi_1]$ ), and to a *quite poor fidelity* for the bottom force (when compared to  $E[\varphi_2]$ ).

As this will be used in the following, let us mention that we can quantify how well  $\hat{\varphi}$  reproduces the behaviour of  $\varphi$  by using the Leave-One-Out cross validation (LOO) error. It is established as follows. For each sample  $\theta^* \in \mathcal{E}$ , a PCE model is established based on

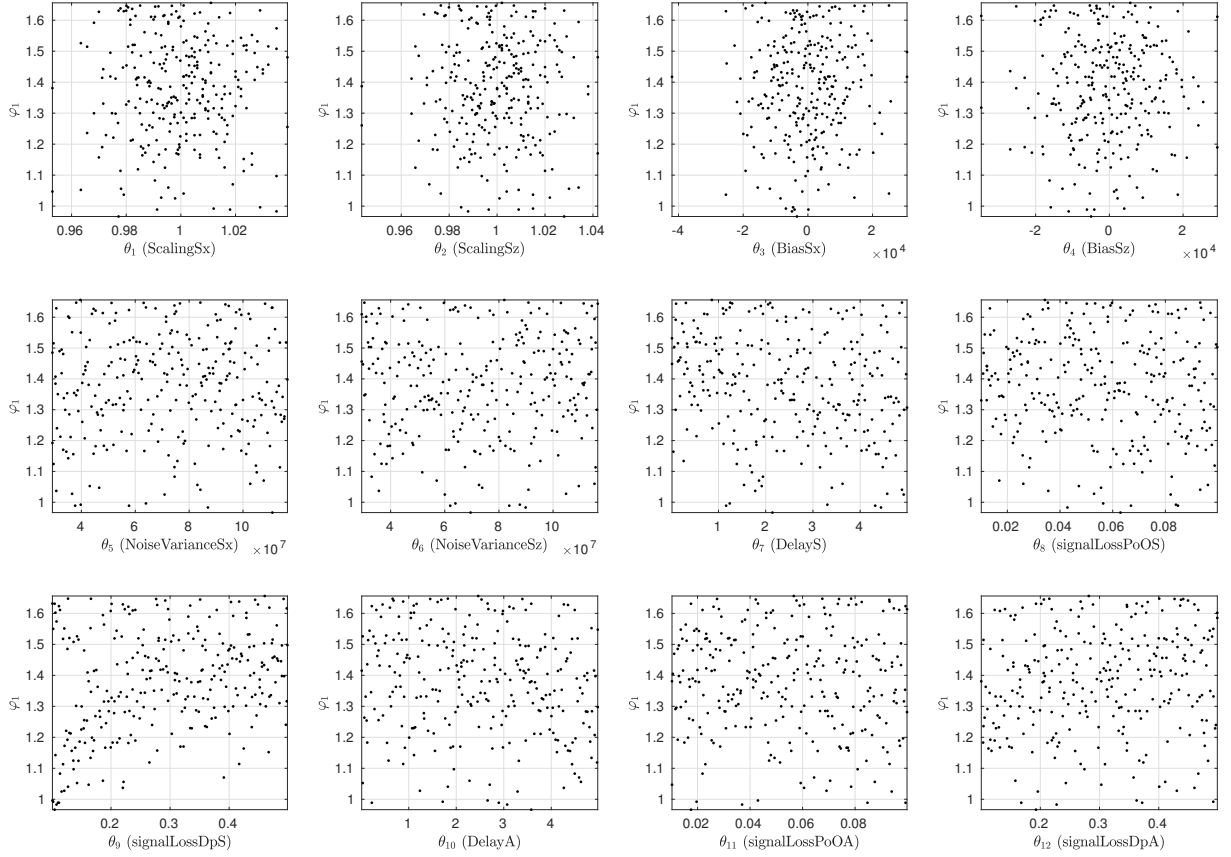


Figure 8: Scatter diagrams showing the value of  $\varphi_1$  (fidelity indicator based on the top velocity of the line), as a function of the twelve parameters describing the artefacts. The dots correspond to 208 samples of  $\Theta$  obtained by Latin Hypercube Sampling (set denoted  $\mathcal{E}$  in the text).

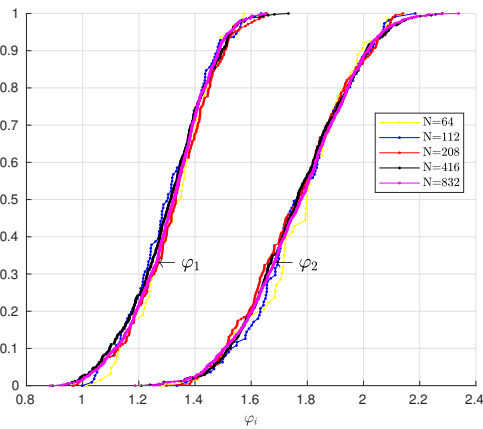


Figure 9: Cumulative distribution functions of  $\varphi_1$  and  $\varphi_2$  obtained from sets  $\mathcal{E}$  of different sizes.

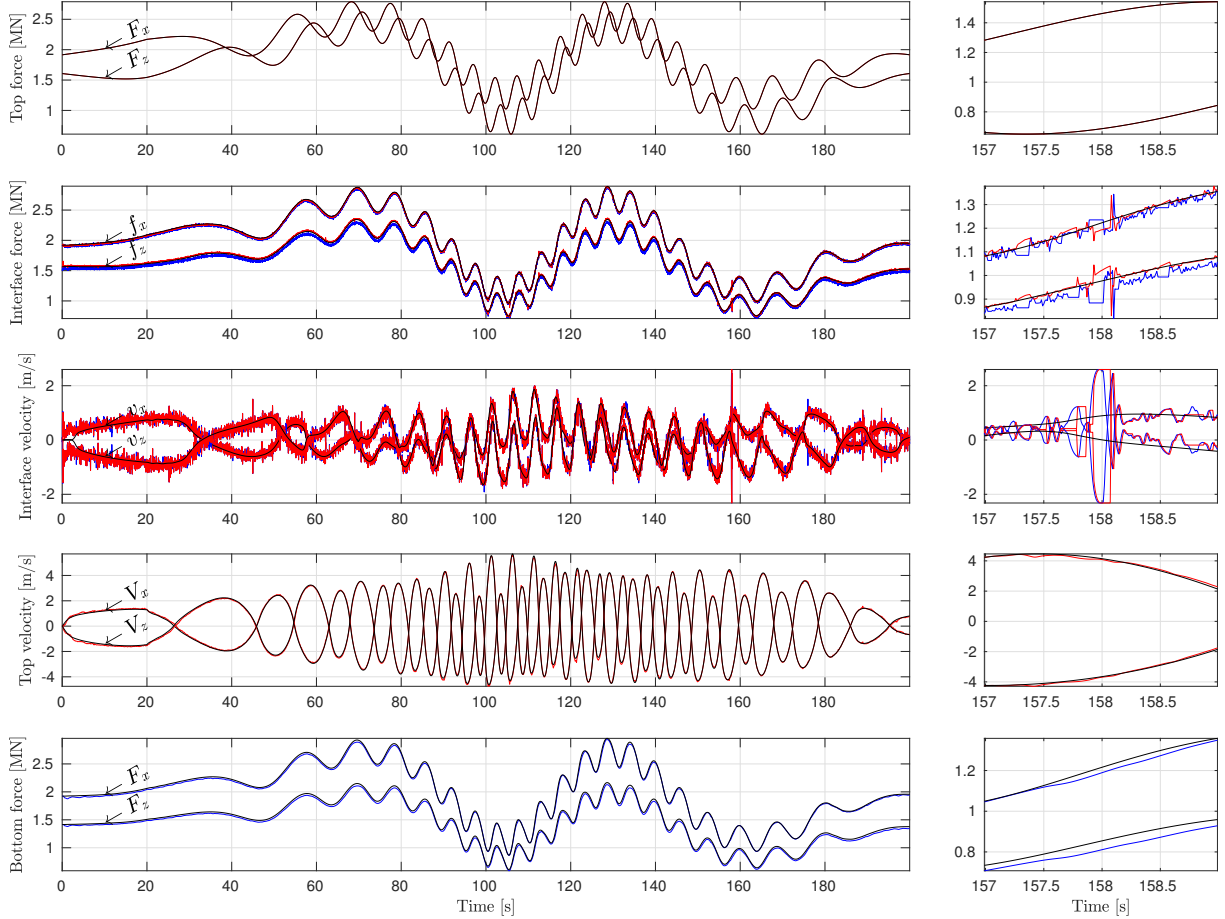


Figure 10: Co-simulation of active truncation with a set of the artefacts leading to the *median* value of  $\varphi_1$ . For this realization, the measurement of  $f_x$  (resp.  $f_z$ ) is affected by a -0.3% (resp. -3%) calibration error, a -0.012 N (resp. 0.28 N) bias, and noise with a standard deviation of 0.040 N (resp. 0.037N), in model scale. The force measurement is delayed by 2.6ms, and has a probability of signal loss of 7.5%, with a duration parameter of 0.47, which corresponds to frequent and short periods of signal loss. On the actuation side, the delay is 1.3ms, and the probability of occurrence and duration parameter of signal loss are 6.8 % and 0.17, respectively. The resulting fidelity indicators are  $\varphi_1=1.33$  and  $\varphi_2=1.62$ .

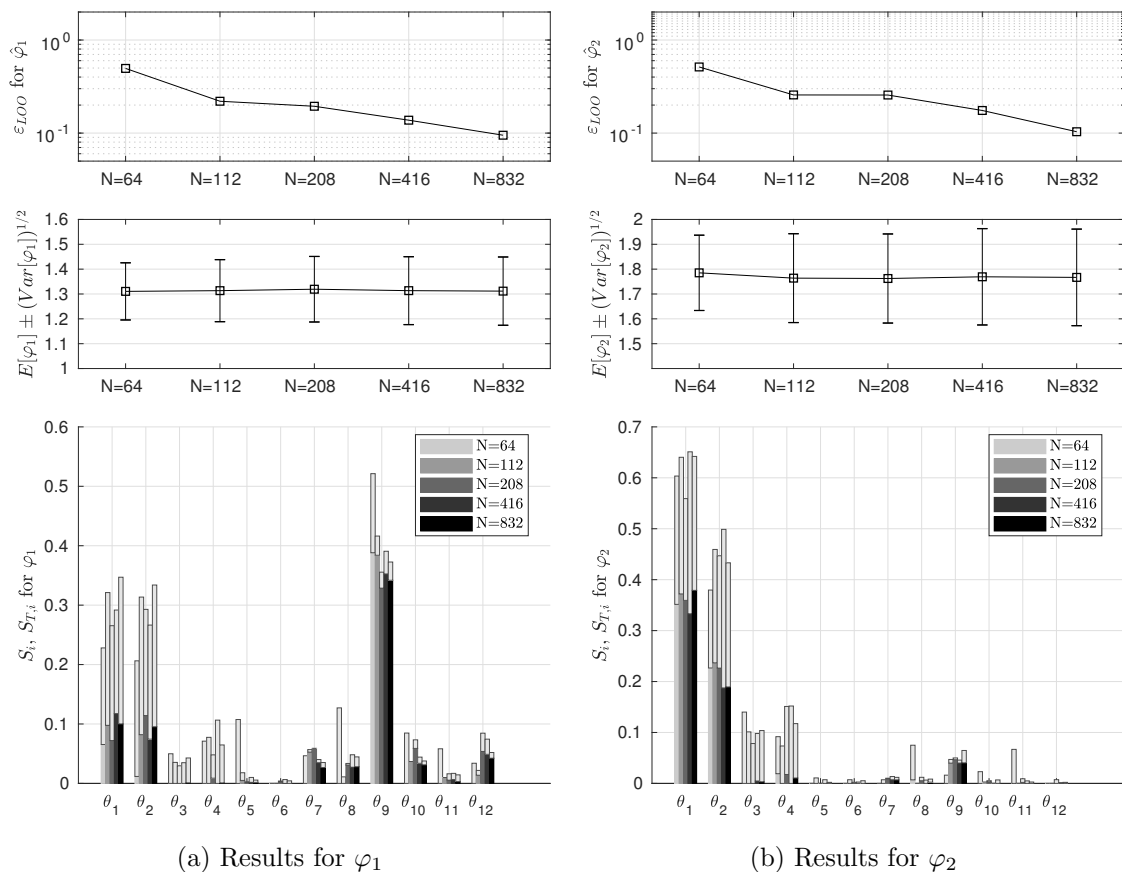


Figure 11: Upper plot: normalized Leave-One-Out cross-validation error for the PCE metamodel of  $\varphi_i$ . Middle plot: expected value and variance of  $\varphi_i$  estimated from initial empirical designs of five different sizes. Lower plot: first-order Sobol' indices of  $\varphi_i$  estimated from these initial experimental designs. The corresponding *total* Sobol' index of each  $\theta_i$  is plotted in grey in the background.

the set  $\mathcal{E} \setminus \theta^*$ , and the values of  $\varphi(\theta^*)$  and  $\hat{\varphi}(\theta^*)$  are compared. Even if this means that, in principle, as many PCE models should be established as elements in  $\mathcal{E}$ , the LOO error can under some circumstances be estimated directly from the single PCE model established from the whole set  $\mathcal{E}$  Marelli and Sudret (2017). The LOO error (normalized by the variance of  $\hat{\varphi}$ ) is usually deemed satisfactory for uncertainty propagation and sensitivity analyses, when it does not exceed  $10^{-1}$  Le Gratiet et al. (2015). This will be commented on later on.

## 4.2 Sensitivity analysis

In some cases, visual inspection of scatter diagrams such as Figure 8, allows one to determine directly which artefact component(s) affects the most the fidelity. This becomes however more difficult for increasing values of  $M$  (the dimension of  $\theta$ ), and particular for the present case with  $M = 12$ . As introduced in Section 2, Sobol' sensitivity indices can instead be used, which are directly deduced from  $\hat{\varphi}$ . Before looking at the Sobol' indices, let us recall

that the *absolute* values of the total Sobol' indices  $S_{T,i}$  are of secondary importance: the  $S_{T,i}$  should be *compared to each other* to identify the most influencing artefacts' parameters. Furthermore,  $S_{T,i}$  can be compared to the first order Sobol' index  $S_i$ , to understand whether the artefact parameter  $\theta_i$  influences the variance of  $\varphi$  alone, or in an interaction with another parameter  $\theta_j$ , or several others.

With these interpretation keys in mind, let us consider the bottom plots in Figure 11, showing the  $S_{T,i}$  and  $S_i$ , estimated from various sizes (or cardinality)  $card(\mathcal{E})$  of  $\mathcal{E}$ . It is seen that for the present problem, reliable insight into the main mechanisms of sensitivity can be obtained for  $card(\mathcal{E})=208$ . If  $card(\mathcal{E})=416$ , finer conclusions can be made regarding the sensitivity to less important parameters, which do not change when  $card(\mathcal{E})=832$ . For both  $card(\mathcal{E})=208$  and 416, the estimated statistical moments are within 1% of the value obtained with for  $card(\mathcal{E})=832$  samples. Note that the recommended values of the LOO error in Le Gratiet et al. (2015) are rather conservative for the present situation, since good convergence of the statistical moments and meaningful sensitivity information are obtained, in spite of an LOO error exceeding 0.1.

Let us first outline the main conclusions that can be drawn from the total Sobol' indices  $S_{T,i}$ , represented by grey bars in Figure 11b (consider for example  $card(\mathcal{E})=416$ ). The fidelity indicator based on the *top velocity* response,  $\varphi_1$ , is very sensitive to  $\theta_9$  (the duration of the signal loss on the force signal) and to the calibration errors of the  $f_x$  and  $f_z$  measurement ( $\theta_1$  and  $\theta_2$ ).  $\varphi_1$  is much less sensitive to the other  $\theta_i$ , and clearly insensitive to noise (described by  $\theta_5$  and  $\theta_6$ ). Focusing now on the *bottom force*, we see that  $\varphi_2$  is mostly sensitive to  $\theta_1$ , then  $\theta_2$  (calibration errors), and then to a much less extent to the biases  $\theta_3$  and  $\theta_4$ , which have both comparable total Sobol' indices.  $\varphi_2$  is slightly sensitive to  $\theta_9$ , the duration parameter for signal loss on the force measurement, and insensitive to the other  $\theta_i$  parameters. We will now relate these results, obtained by a systematic approach, to their physical causes.

**The effect of white noise** It is clear from Figure 10 that the noise affecting force measurements (parametrized by  $\theta_5$  and  $\theta_6$ ) induces a significant velocity response at the truncation point. This response is however filtered mechanically by drag and, to a less extent, structural damping, before reaching the top and bottom of the mooring line. Therefore noise does not significantly affect the fidelity indicators  $\varphi_1$  and  $\varphi_2$ . The fact that the  $S_{T,i}$  associated to this artefact are negligible, means that the corresponding parameters  $\theta_5$  and  $\theta_6$  (noise variances) could have been set to deterministic values (here, zero), without affecting the variance of  $\varphi$ .

**Signal loss** A natural question when looking at Fig. 11a, is why the top velocity (or  $\varphi_1$ ) is more sensitive to signal loss, when it acts on the force sensor (duration parameter  $\theta_9$ ) rather than when it acts on the velocity actuation (parameter  $\theta_{12}$ ). Indeed, the force sensors feeds the numerical substructure, while the actuator controls the bottom part of the

physical substructure, whose response *directly* enters in the definition (7) of  $\varphi_1$ . The reason is the following. When signal loss on the velocity command happens, the velocity of the truncation point keeps a *constant* value. On the other hand, signal loss on the force sensor may cause *large variations* of the truncation point's velocity, due to the interaction with the numerical substructure that was commented in detail in Section 3 and seen in Figure 5. Both the *amplitude* of these perturbations and their *duration* increase when the signal loss characteristic duration increases, which enhances their propagation to the top of the mooring line.

**Effect of the anisotropic properties of the mooring** The fact that  $\varphi_2$  is more sensitive to  $\theta_1$  (calibration error for  $f_x$  measurement) than to its counterpart  $\theta_2$  (acting on  $f_z$ ) can be explained as follows. Transverse motions of the mooring line are subjected to drag damping forces, while axial motions are only damped by structural damping, which means that, with the present choice of  $\alpha_2$  and the present frequency range of motions, transverse motions will be subjected to a significantly higher level of damping than axial motions. Consequently, an axial dynamic force error will be less damped than its transverse counterpart. Since the mooring line forms an angle of  $\gamma = 39.2^\circ$  with respect to the  $x$ -axis at the truncation point, the axial forces have an  $x$ -component larger than their  $z$ - component, and a calibration error on  $f_x$  (parametrized by  $\theta_1$ ) will play a greater role for  $\varphi_2$  than a calibration error on  $f_z$  (parametrized by  $\theta_2$ ).

Also, as explained earlier, Total Sobol' indices and first-order indices differ when there is an interaction between two (or more)  $\theta_i$ . The nature of this interaction can be determined by considering higher-order Sobol' indices (not shown here). Note that in principle, a finer PCE model (with lower LOO error) would be needed to obtain accurate estimates of the higher-order Sobol' indices, so only trends will be commented here. We found for example that the interaction between  $\theta_1$  and  $\theta_2$  explains  $\approx 20\%$  of the variance of  $\varphi_1$ , and  $\approx 15\%$  of the variance of  $\varphi_2$ . This is due to the fact that if  $\theta_1$  and  $\theta_2$  differ significantly from each other, the *direction* of the force at the truncation point will be affected. Since the stiffness and damping properties of the line are not isotropic, as explained earlier, this change in *direction* will have a significant effect on the fidelity.

**Effect of the bias** From Figure 11, we see that biases have a significant influence on  $\varphi_1$  and  $\varphi_2$  (total Sobol indices), and that this influence is due to interactions ( $S_5 \ll S_{T,5}$  and  $S_6 \ll S_{T,6}$  in both cases). Here, the mechanisms in play are slightly different for  $\varphi_1$  and  $\varphi_2$ .

By again studying higher-order Sobol indices, it can be shown that the interaction between  $\theta_1$  and  $\theta_3$  (resp.  $\theta_2$  and  $\theta_4$ ) explain  $\approx 10\%$  of the variance of  $\varphi_2$ . This interaction is induced by the *pretension* at the truncation point, denoted  $T_0^*$ . Indeed, for example, when a scaling error  $\theta_1$  affects  $f_x$  only, it is equivalent to a bias of  $(\theta_1 - 1)T_0^* \cos \gamma$  being added to  $\theta_3$ , and transferred to the anchor point. Coupling terms between  $\theta_1$  and  $\theta_3$  will therefore be

generated in the Sobol' decomposition (5) of  $\varphi_2$ , due to the logarithm in (8).

Biases should, in principle have little influence on  $\varphi_1$ , since constant force will simply lead to a constant offset, and not change the (linear) dynamical properties of our substructures. However, about  $\approx 10\%$  of the variance of  $\varphi_1$  is due to one-to-one interactions between  $\theta_1$ ,  $\theta_2$ ,  $\theta_3$ , and  $\theta_4$ . This is due to the following effect. In the horizontal direction, for  $t < 0$ , the component of the pretension is  $T_0^* \cos \gamma$  at the truncation point. For  $t \geq 0$ , it suddenly changes to  $\theta_3 + (\theta_1 - 1)T_0^* \cos \gamma$ , when artefacts are applied to the force signal. This impulsive load causes a transient response, visible in Figure 10, which has a minor, but noticeable, influence on  $\varphi_1$ .

## 5 Conclusion

In the present paper, we considered active truncation as an alternative technique to perform model testing of ultra-deep water floating systems in existing ocean laboratories. We assessed the performance of an active truncation setup through its associated fidelity. We showed how the fidelity could be jeopardized by multiple, heterogeneous and random artefacts, originating from the control system (including sensors, actuators and controllers) that connects the numerical and physical parts of the setup. We outlined a method to evaluate the expected fidelity of the setup, and its variability due to the uncertainties on the artefacts. Finally, a *systematic* analysis method based on Sobol' indices allowed us to determine the sensitivity of the fidelity to each of the involved artefacts. This latter result provides valuable and objective indications to improve fidelity in an operational context. Using the polynomial chaos expansions of the fidelity made this sensitivity analysis possible at a reasonable computational cost.

A case study addressing a taut polyester mooring line allowed to gain insight in the complex mechanisms taking place in active truncation, combining the dynamics of the slender structures and the imperfect coupling at the truncation point. A total of ten artefacts were included in the analysis, and the importance of calibration errors and signal loss at the force sensors level was put in evidence. It must be emphasized that, since the fidelity is a nonlinear function of the artefacts parameters, these conclusions are valid for the present system and set of artefacts only.

The present method is currently extended to a complete framework, which also allows to (1) determine the feasibility of an active truncation test, by evaluating its probability of failure due to too low fidelity, and (2) determine the corresponding admissible bounds on the artefact parameters.

## Acknowledgment

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## A Verification of `fixedFreeCableSegment`

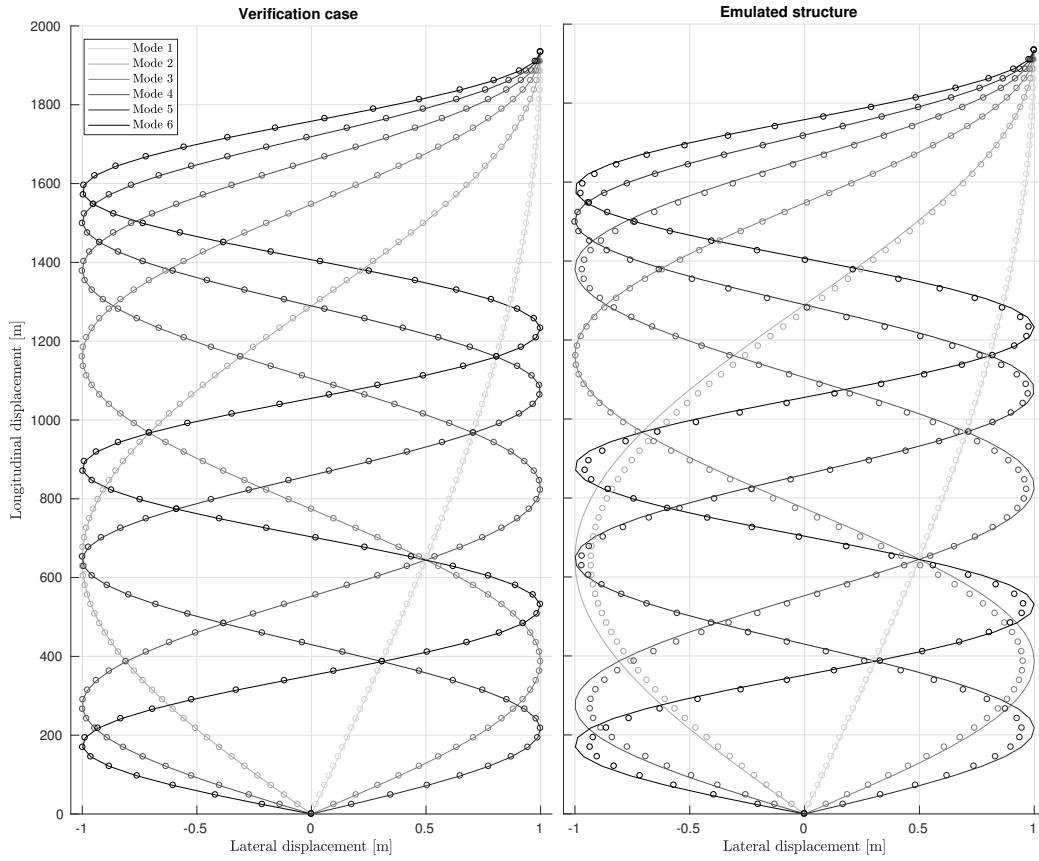
As verification of the FE implementation, the eigenvalues and associated modeshapes computed from `fixedFreeCableStructure` are compared to a known analytical solution. We consider a polyester line, commonly used in mooring systems, whose nominal properties are given in Table 1. Its length is  $L = 1934$  m and it is subjected to a vertical top tension of  $T_0 = 2.5$  MN. Under the assumptions of infinite axial stiffness and zero submerged weight, its eigenfrequencies  $\omega_i$  and associated modeshapes  $\phi_i$  are given by:

$$\forall i \in \mathbb{N}^*, \omega_i = \frac{(2i-1)\pi}{2L} \sqrt{\frac{T_0}{m}} \text{ and } \phi_i(z) = (-1)^{i+1} \sin\left(\frac{(2i-1)\pi z}{2L}\right) \quad (9)$$

where  $m$  is the mass and added mass per unit of length, and  $\phi_i$  is normalized so that  $\phi_i(L) = 1$ . The six first modeshapes obtained from this analytical solution are represented with solid lines in Fig. 12, and the values of the 15 first eigenperiods are tabulated in the second column of Table 3. These eigenmodes are compared to the results obtained for a `fixedFreeCableSegment` object with  $n_{el}=500$  elements, whose axial stiffness has been increased by one order of magnitude (to mimic the infinite stiffness assumption used in the analytical model), and whose submerged weight has been set to zero. The eigenmodes of a `fixedFreeCableSegment` object are obtained numerically from the eigenvalue analysis of  $M^{-1}K$ , where  $M$  and  $K$  are obtained from the nonlinear static analysis. The corresponding eigenperiods are tabulated in the third column of Table 3, and the difference with the analytical solution is found to be insignificant. When  $n_{el}$  is decreased to 80 elements (fourth column of Table 3), the error is less than 1% for the 13 first modes, and the first modeshapes, compared in Fig. 12a, also show excellent agreement. For higher modes, with eigenperiods less than 1.80 s, the model with  $n_{el}=80$  becomes too coarse, with less than 12 elements per wavelength  $4L/(2i-1)$ , and the estimated eigenperiods become erroneous. So provided that  $n_{el}$  is chosen adequately, the dynamic system modelled by `fixedFreeCableSegment` can be considered as verified against the corresponding analytical solution.

In reality, several physical effects will make the eigenmodes of a a polyester line deviate from the ideal solution (9). (1) First, the elasticity of the polyester somewhat influences the dynamics of long lines. This is shown in the fifth column of Table 3, in which eigenperiods are evaluated from a vertical `fixedFreeCableSegment`, now featuring its nominal stiffness. While the elasticity of the line does not influence significantly the ten first transverse modes, it must be accounted for when higher modes (with associated eigenperiods lower than 2.41s, in the present case) need to be modelled. (2) The submerged weight of the slender structure causes tension variations throughout the water column, which also affects the eigenmodes. By considering the the sixth column of Table 3, it is seen that this effect has an impact on all modes, including the the lower modes, making the corresponding eigenperiods deviate by 2 to 3% from the previous solution. (3) Then, since such a polyester line is in general installed in an oblique way, it will exhibit static lateral deflections (of the order of 1% of the

structure's length in the present case), due to its submerged weight. As shown in the sixth column of Table 3, this change of static configuration has some effect on all eigenmodes. (4) Finally, the oblique line is subjected to the shear current introduced in the previous section. It is found to have an insignificant additional effect on the eigenmodes (last column in Table 3). Note however that current has an important effect on the drag-induced damping of transverse motions.



(a) Verification case: ideal cable.

(b) Modal analysis of the emulated structure with nominal weight in water, stiffness, and under inclined top force and current loads, and comparison with an ideal cable.

Figure 12: Modeshapes corresponding to the six first eigenmodes of the fixed-free cable structure. Corresponding eigenperiods can be found in Table 3. In both figures, the analytical solutions for an ideal (weightless and infinitely stiff) cable are plotted with solid lines, and numerical results using the `fixedFreeCableSegment` class with 80 elements are plotted with circle markers.

Table 3: Eigenperiods in seconds corresponding to the 15 first modes of a 1934m long cable subjected to a top tension of 2.5MN. In italic: deviation in percents between the analytical solution (transverse vibrations of a weightless and infinitely stiff string) and various numerical solutions computed with the `fixedFreeCableSegment` class.

	Analytical solution	Numerical solution using the <code>fixedFreeCableSegment</code> class											
		10*nominal	10*nominal	Nominal	Nominal	Nominal	Nominal	Nominal	Nominal	Nominal	Nominal		
Axial stiffness	Infinite	Weightless	Weightless	Weightless	Weightless	Weightless	Weightless	Weightless	Weightless	Weightless	Weightless		
Weight in water	Weightless	Weightless	Weightless	Weightless	Weightless	Weightless	Weightless	Weightless	Weightless	Weightless	Weightless		
Top force direction	-	Vertical	Vertical	Vertical	Vertical	Vertical	Vertical	Vertical	Vertical	Vertical	Vertical		
Current	-	None	None	None	None	None	None	None	None	None	None		
Number of elements	-	500	80	80	80	80	80	80	80	80	80		
Mode 1	49.15	49.15	<i>0.00</i>	49.15	<i>0.00</i>	49.39	<i>0.48</i>	50.68	<i>3.12</i>	50.17	<i>2.08</i>	50.17	<i>2.07</i>
Mode 2	16.38	16.38	<i>0.00</i>	16.38	<i>-0.01</i>	16.46	<i>0.47</i>	16.78	<i>2.42</i>	16.66	<i>1.71</i>	16.66	<i>1.71</i>
Mode 3	9.83	9.83	<i>0.00</i>	9.83	<i>-0.04</i>	9.87	<i>0.44</i>	10.06	<i>2.33</i>	9.99	<i>1.62</i>	9.99	<i>1.62</i>
Mode 4	7.02	7.02	<i>0.00</i>	7.02	<i>-0.08</i>	7.05	<i>0.41</i>	7.18	<i>2.28</i>	7.13	<i>1.57</i>	7.13	<i>1.57</i>
Mode 5	5.46	5.46	<i>0.00</i>	5.45	<i>-0.13</i>	5.48	<i>0.35</i>	5.58	<i>2.22</i>	5.54	<i>1.51</i>	5.54	<i>1.51</i>
Mode 6	4.47	4.47	<i>0.00</i>	4.46	<i>-0.19</i>	4.48	<i>0.29</i>	4.56	<i>2.15</i>	4.53	<i>1.45</i>	4.53	<i>1.44</i>
Mode 7	3.78	3.78	<i>-0.01</i>	3.77	<i>-0.27</i>	3.79	<i>0.21</i>	3.86	<i>2.07</i>	3.83	<i>1.37</i>	3.83	<i>1.36</i>
Mode 8	3.28	3.28	<i>-0.01</i>	3.26	<i>-0.36</i>	3.28	<i>0.12</i>	3.34	<i>1.98</i>	3.32	<i>1.27</i>	3.32	<i>1.27</i>
Mode 9	2.89	2.89	<i>-0.01</i>	2.88	<i>-0.46</i>	2.89	<i>0.02</i>	2.95	<i>1.87</i>	2.92	<i>1.17</i>	2.92	<i>1.16</i>
Mode 10	2.59	2.59	<i>-0.01</i>	2.57	<i>-0.58</i>	2.58	<i>-0.10</i>	2.63	<i>1.76</i>	2.61	<i>1.05</i>	2.61	<i>1.05</i>
Mode 11	2.34	2.34	<i>-0.02</i>	2.32	<i>-0.70</i>	2.41	<i>2.92</i>	2.41	<i>2.90</i>	2.41	<i>2.88</i>	2.41	<i>2.87</i>
Mode 12	2.14	2.14	<i>-0.02</i>	2.12	<i>-0.84</i>	2.34	<i>9.28</i>	2.38	<i>11.30</i>	2.36	<i>10.53</i>	2.36	<i>10.53</i>
Mode 13	1.97	1.97	<i>-0.02</i>	1.95	<i>-0.99</i>	2.13	<i>8.30</i>	2.17	<i>10.31</i>	2.15	<i>9.54</i>	2.15	<i>9.54</i>
Mode 14	1.82	1.82	<i>-0.03</i>	1.80	<i>-1.16</i>	1.96	<i>7.44</i>	1.99	<i>9.43</i>	1.98	<i>8.67</i>	1.98	<i>8.67</i>
Mode 15	1.69	1.69	<i>-0.03</i>	1.67	<i>-1.34</i>	1.81	<i>6.67</i>	1.84	<i>8.65</i>	1.83	<i>7.90</i>	1.83	<i>7.89</i>

## B Convergence and parameter study for the co-simulation

We consider and the taut polyester mooring line, truncated with  $\alpha=0.8$ , and exposed to the current and wave-induced loads described in Section 2. The response of the substructured system is evaluated by the co-simulation procedure outlined previously. Each of the four parameters is varied, keeping the other ones constant and equal to the following nominal values:  $n_{el}=80$  elements,  $\delta t=10$  ms,  $\epsilon_v=10^{-6}$  m/s and  $\epsilon_f=0.1$  N. As discussed in the previous section, the QoI for our problem are the top velocity of  $\mathbf{p}$  and the bottom force of  $\mathbf{n}$ . The following indicators are therefore used to study convergence:

$$\epsilon_1 = \left( \frac{\int_0^T (V_{x,top}(t) - V_{x,top}^\infty(t))^2 dt}{\int_0^T V_{x,top}^\infty(t)^2 dt} + \frac{\int_0^T (V_{z,top}(t) - V_{z,top}^\infty(t))^2 dt}{\int_0^T V_{z,top}^\infty(t)^2 dt} \right)^{1/2} \quad (10)$$

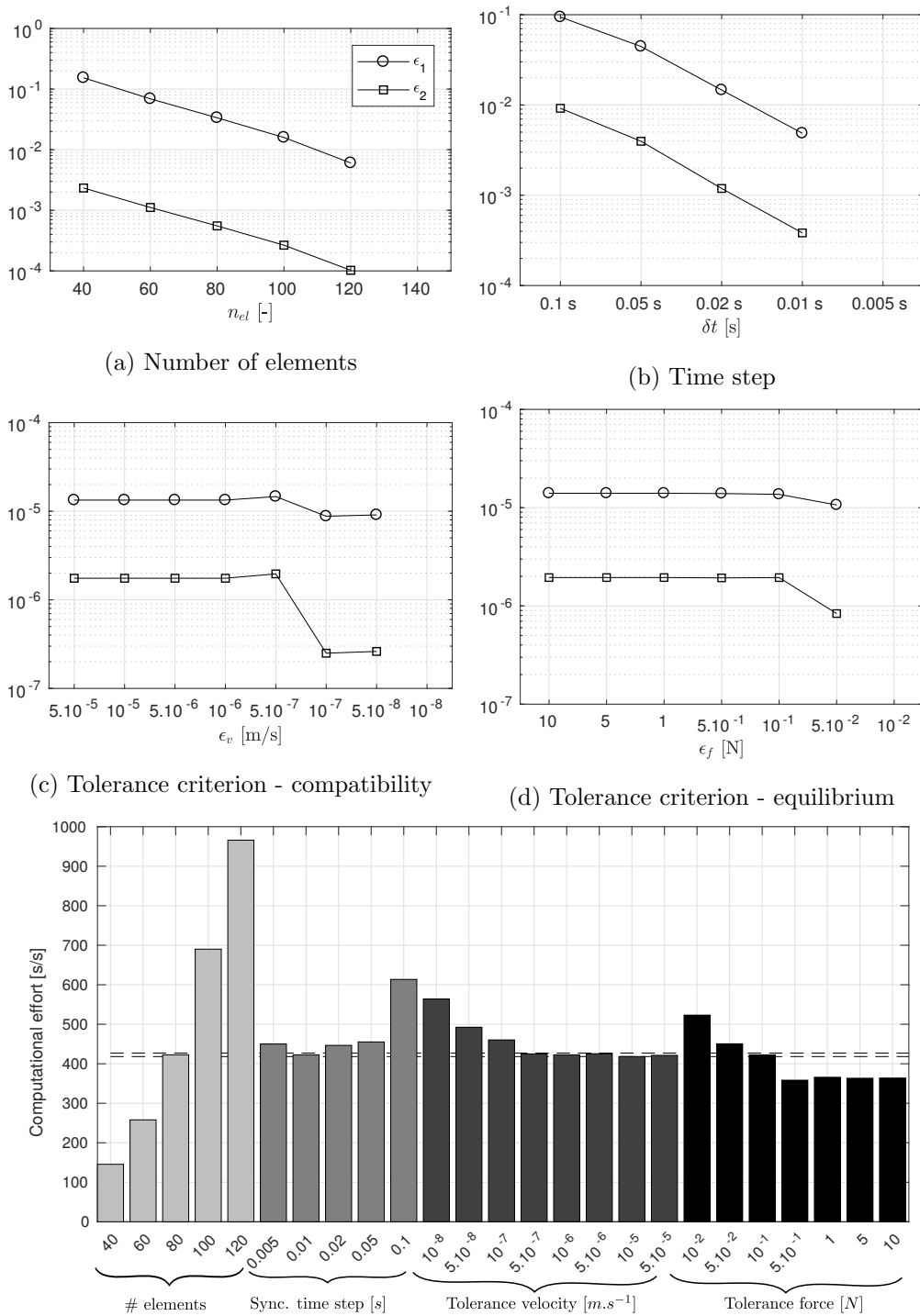
$$\epsilon_2 = \left( \frac{\int_0^T (F_{x,bottom}(t) - F_{x,bottom}^\infty(t))^2 dt}{\int_0^T F_{x,bottom}^\infty(t)^2 dt} + \frac{\int_0^T (F_{z,bottom}(t) - F_{z,bottom}^\infty(t))^2 dt}{\int_0^T F_{z,bottom}^\infty(t)^2 dt} \right)^{1/2} \quad (11)$$

where the  $^\infty$  superscript refers to the time series obtained with the finest mesh, smallest synchronization time step or tolerance value, depending on which parameter is varied. Fig. 13a to 13d show the variations of  $\epsilon_1$  and  $\epsilon_2$  as a function of each parameter, and Fig. 13e shows the effect of the parameters on the computational time.

As expected,  $\epsilon_1$  and  $\epsilon_2$  decrease when refining the mesh (Fig. 13a), while the computational time increases proportionally to  $n_{el}^2$  (Fig. 13e). As seen in Section 4, the present study requires a possibly large number of co-simulations,  $n_{el}=80$  is selected, which allows keeping computational costs to an acceptable level, with an  $\epsilon_1$  error of the order of 2%.

Convergence is also clearly observed when the synchronization time step is reduced (Fig. 13b). It can be observed (Fig. 13e) that the computational time is minimum for  $\delta t=10$  ms, and increases significantly when  $\delta t=100$ ms. Indeed, even if reducing total number of synchronizations during the given simulation time, increasing  $\delta t$  leads to a larger required number of iterations (lines 6-14 in Alg. 1) at each synchronization step. On the other hand, it can be observed that the computational time is larger for  $\delta t=5$ ms than for  $\delta t=10$ ms. In that case, even if very few iterations are required to achieve compatibility and equilibrium, the total computational burden increases due to some expensive operations (such as writing data), which are performed at the end of each synchronization step.

Finally, as expected when considering line 11 in Alg. 1,  $\epsilon_v$  and  $\epsilon_f$  play a symmetric role. For a given  $\epsilon_f$  for example, decreasing  $\epsilon_v$  will only have an influence on the outcome of the co-simulation (and thus on  $\epsilon_1$  and  $\epsilon_2$ ) if it is  $\epsilon_v$ , and not  $\epsilon_f$ , that forces the iteration process to continue. Indeed, when  $\epsilon_v$  is chosen to be very large, the dynamic equilibrium condition will be the limiting constraint, and the value of  $\epsilon_f$  will thus steer the number of iterations. When  $\epsilon_v$  is decreased and reaches a certain threshold, which depends on the mechanical impedance of the structure, it may be either the equilibrium or the compatibility condition



(e) Computational time: number of seconds required per simulated second. The dashed lines correspond to the mean value  $\pm$  one standard deviation of the duration of the simulation using the nominal set of parameters.

Figure 13: Convergence study. Effect of varying the number of elements, the synchronization time step and synchronization tolerances on the error indicators  $\epsilon_1$  and  $\epsilon_2$  (four top figures), and on the computational time (lower figure).

that steers the number of iterations, at a given synchronization step. Finally, decreasing further  $\epsilon_v$  will enforce an increased accuracy on the compatibility condition, which decreases the error, and increases the number of iterations and the computational time. This shift is clearly happening for  $\epsilon_v=10^{-7}$  m/s in Fig. 13c. It should however be noted, that within the range of investigated  $\epsilon_v$  and  $\epsilon_f$ , the errors  $\epsilon_1$  and  $\epsilon_2$  are extremely small.