

Modeling and Identification of the Restoring Force of a Marine Riser

L. Torres* C. Verde* G. Besançon** J.D. Avilés***

* *Instituto de Ingeniería-UNAM, Mexico City, Mexico (e-mail: ftorreso@iingen.unam.mx, verde@unam.mx)*

** *Control Systems Department, Gipsa-lab, Ense³, Grenoble INP, Saint-Martin d'Hères, France, and Institut Universitaire de France (e-mail: Gildas.Besancon@Grenoble-inp.fr)*

*** *Facultad de Ingeniería-UNAM, Mexico City, Mexico (e-mail: jdav@fi-b.unam.mx)*

Abstract: The purpose of this paper is to provide a novel model to characterize the nonlinear restoring force of a marine vertical riser by using position-and-velocity-dependent polynomials. This model permits the obtaining of a specific state space representation—via the Liénard transformation—for the design of state observers that identify the structural parameters of vertical risers. The main results presented here are: (i) an approximation of the nonlinear restoring force by means of polynomials and its incorporation into a distributed parameter (DP) model, (ii) the transformation of the DP model into a Liénard system and (iii) an analysis of its observability and identifiability properties.

Keywords: Marine systems, pipelines, structural parameters, parameter identification, state observers

1. INTRODUCTION

The marine risers are underwater pipelines that perform large displacements and vibrations because of external forces. Risers adopt an important role in the extraction of petroleum from the sea (Lee (2009)) as the connection between a platform and an oil wellhead placed on the seabed. Their main use is the transportation of the crude oil or sludge when a well is drilled; additionally, they can be used to safeguard the drilling column. Dynamic behavior of a riser can be modeled numerically as a harmonic oscillator with distributed parameters (masses, springs and dampers) along its structure, which is in contact with several outward forces (ocean currents, waves, platform motion) that determine its behavior over time (Niedzwecki and Liagre (2003), Furnes (2000)).

If a structure is excited by external forces with frequencies near its natural frequency, the structure vibration is amplified rendering the entire system at risk of becoming unstable. This phenomenon is known as resonance. In mechanical structures, vibrations cause wear and can produce anomalies with undesirable outcomes. Therefore, vibration and resonance are widely studied phenomena, particularly in the reliability and assessment of building constructions Doebling (1996).

Vibration is present during the exploration and exploitation processes of petroleum in deep waters. This is caused by the force exerted by ocean currents, vortexes, the waves moving the platform and the wind on the oil extraction structures. Riser vibration induces mechanical stress, pro-

blems of fatigue and crack propagation, which require expensive inspections and repairs. Furthermore, the majority of the crude oil fields is localized in zones prone to extreme weather such as hurricanes, cyclones, polar storms, etc. Thus, the installations are always susceptible to structural damages.

To avoid economic losses and environmental damage, considering automatic on-line monitoring systems is necessary in order to estimate structural changes in the risers, i.e., structural health monitoring systems (SHMs). Several methodologies have been developed to monitor the structural conditions of the risers; one of these is the use of dynamic data—obtained by acceleration and vibration measurement techniques—which continuously update the parameters of a structural model. Among these works one finds Ghanem and Shinozuka (1995), Shinozuka and Ghanem (1995) and Doebling (1996), which are widely used—mainly in the case of land structures—to treat the parameter identification in a linear dynamical context. In the case of marine risers, however, the parameter identification problem must be tackled with nonlinear techniques, because these systems have a strong nonlinear behavior because of large displacements of the floating system caused by the environmental loads. In general, there are several factors that determine the nonlinear behavior of a riser modeled by a nonlinear restoring force: (i) the oil flowing inside the riser and its interaction with the inner walls which have a constitution that is variant according to the building materials, (ii) the interaction of the outer walls with the sea and (iii) the vibration induced vortexes (VIV).

* The authors acknowledge the financial support from DGAPA - UNAM and CONACYT (Consejo Nacional de Ciencia y Tecnología).

In this regard, many identification techniques have been developed for the identification of nonlinear systems, which can be classified according to seven categories: linearization-based methods, time-domain and frequency domain methods, modal methods, time-frequency analysis, black-box modeling and structural model updating. The reader may consult Kerschen et al. (2006) for an overview of their application on structural system identification. Another option to identify parameters of nonlinear systems (in the time domain) is the use of state observers (Busvelle and Gauthier (2002); Jiang et al. (2004); Besançon and Ticlea (2007)), which have already been employed in structural parameter identification, e.g., Lin and Betti (2004), Garrido et al. (2004), Angeles and Alvarez-Icaza (2005) and Jiménez-Fabián and Alvarez-Icaza (2010).

Various works have proposed models for the nonlinear restoring force of marine risers, e.g., force-decomposition models (Sarpkaya, 2004), single-degree-of-freedom (SDOF) models (Basu and Vickery, 1983), and wake-body coupled models. In Panneer-Selvam and Bhattacharyya (2001), the authors developed an iterative scheme for the identification of the hydrodynamic coefficients in a Morison type model and included in their analysis a nonlinear stiffness parameter (Duffing coefficient). In Bishop and Hassan (1964), the authors suggested the use of a Van der Pol oscillator to describe the time-varying forces. In Violette et al. (2007), a weak-oscillator model was developed, which is attached to each node (of the discrete solution) to simulate the hydrodynamic force in cross-flow direction.

The objective of this paper is to obtain a generalized model of the nonlinear restoring force of a marine riser using a pair of polynomials that depend on the velocity and position variables; the model then provides the following advantages: (i) the resulting model of the riser, a nonlinear polynomial oscillator, can be put into a Liénard representation to design state observers, and (ii) choosing high order polynomials to approximate nonlinearities with non-polynomial nature is possible. Finally, another contribution in this article is the analysis of observability/identifiability of the finite version of the structural DP model presented in Niedzwecki and Liagre (2003) and modified here with the inclusion of the nonlinear polynomial restoring force. The finite version in space is achieved by employing the finite difference method (FDM) and the Liénard transformation.

The paper is organized as follows. Section 2 presents the model used to simulate the behavior of a marine riser, its spatial discretization and its transformation into a Liénard system. Additionally, several expressions are shown for the modeling of the restoring force, waves and the hydrodynamic force. In Section 3, the observability/identifiability issue is discussed. Finally, results of simulation are presented in Section 4, and concluding remarks are drawn in Section 5.

2. RISER PHYSICAL MODEL

In Niedzwecki and Liagre (2003), to estimate the parameters of a marine riser, the authors proposed the following fourth order quasi-linear partial differential equation

(PDE) describing the riser displacements for an external excitation $u(z, t)$:

$$EI \frac{\partial^4 \nu(z, t)}{\partial z^4} - T \frac{\partial^2 \nu(z, t)}{\partial z^2} + m \frac{\partial^2 \nu(z, t)}{\partial t^2} + c \frac{\partial \nu(z, t)}{\partial t} + p(z, t) = u(z, t) \quad (1)$$

where $(z, t) \in (0, L) \times (0, \infty)$ are the time and space coordinates respectively, $\nu(z, t)$ is the horizontal displacement of the marine riser, m is the mass per unit, c is the linear viscous drag coefficient, T is tension, EI is the bending stiffness, and the term $p(z, t)$ represents a nonlinear restoring-damping force related to the nonlinear drag force which depends of time and space. Notice that in model (1), in order to have a simplified and useful model to develop identification approaches, the physical properties as well as the tension are assumed to be uniform along the length of the riser and time invariant. Indeed, this model was used in Niedzwecki and Liagre (2003) to elaborate a frequency domain identification algorithm.

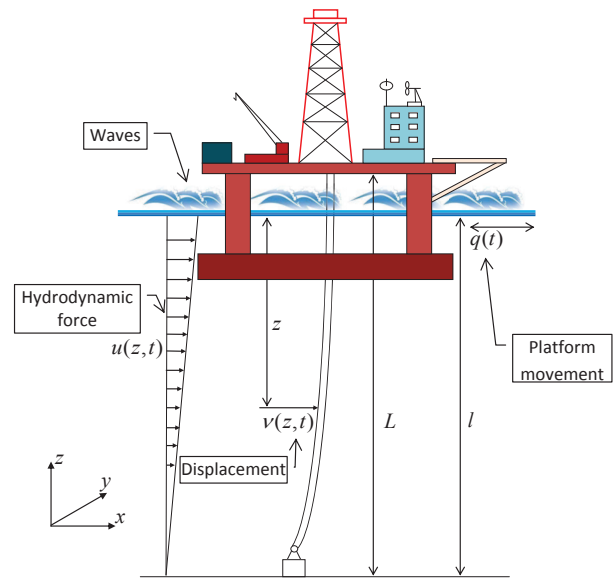


Fig. 1. Schematic of a marine riser.

2.1 Hydrodynamic force

Hydrodynamic force acting on the risers can be calculated using a modified Morison's equation for a cylinder in motion (Morison et al., 1950). Thus, the horizontal hydrodynamic force can be expressed as

$$u(z, t) = \rho_w C_M \frac{\pi D^2}{4} \frac{\partial d(z, t)}{\partial t} + \frac{1}{2} \rho_w C_D D (V(z, t) - d(z, t)) |V(z, t) - d(z, t)|. \quad (2)$$

Coefficients intervening in this equation are C_M , the inertia coefficient and C_D , the drag coefficient. Furthermore, ρ_w is the water density, D is the outer marine riser diameter and $V(z, t)$ is the current velocity, whereas $d(z, t)$ and $\partial d(z, t)/\partial t$ are the velocity and acceleration of the waves, which can be calculated by the following expressions (Wheeler (1970)):

$$d(z, t) = \sum_{j=1}^{\infty} A_j \omega_j \frac{\cosh(k_j z)}{\sinh(k_j l)} \cos(\omega_j t + \phi_j)$$

$$\frac{\partial d(z, t)}{\partial t} = \sum_{j=1}^{\infty} A_j \omega_j \frac{\cosh(k_j z)}{\sinh(k_j l)} \sin(\omega_j t + \phi_j)$$

where A_j is the wave component amplitude obtained based upon a particular random, ω_j is the corresponding wave frequency and ϕ_j is the random phase angle assumed to be uniformly distributed over the interval $[0, 2\pi]$, and k_j is the wave number and is related to ω_j through the linear dispersion relation for a specified water depth l :

$$\omega_j^2 = gk_j \tanh(k_j l).$$

2.2 Nonlinear restoring force

In Niedzwecki and Liagre (2003), the authors suggested that the nonlinear drag force $p(z, t)$ has polynomial types of non-linearities. In particular, they proposed a general non-linear damping-restoring term based upon the combination of classic Duffing and Van der Pol nonlinearities as follows:

$$p(z, t) = k_3 \nu^3(z, t) + \frac{c_3}{3} \dot{\nu}^3(z, t) \quad (3)$$

where k_3 is the Duffing coefficient, c_3 is the Van der Pol coefficient and notation $\dot{\nu}$ stands here for partial derivative with respect to time. This model will also be considered in the present paper as a reference in order to assess the performance of the further proposed nonlinear identification algorithm.

For this identification, a nonlinear restoring force model composed of two polynomials is presented, which depend on the position and velocity variables, with orders η and n respectively:

$$p(z, t) = a_1 \nu(z, t) \dot{\nu}(z, t) + a_2 \nu^2(z, t) \dot{\nu}(z, t) + \dots + a_\eta \nu^\eta(z, t) \dot{\nu}(z, t) + b_1 \nu(z, t) + b_2 \nu^2(z, t) + \dots + b_n \nu^n(z, t). \quad (4)$$

Thus, the complete model (1) with the force (4), i.e.,

$$EI \frac{\partial^4 \nu(z, t)}{\partial z^4} - T \frac{\partial^2 \nu(z, t)}{\partial z^2} + m \ddot{\nu}(z, t) + \underbrace{(c + a_1 \nu(z, t) + a_2 \nu^2(z, t) + \dots + a_\eta \nu^\eta(z, t))}_{F_0(\nu)} \dot{\nu}(z, t) + \underbrace{(b_1 \nu(z, t) + b_2 \nu^2(z, t) + \dots + b_n \nu^n(z, t))}_{G_0(\nu)} = u(z, t) \quad (5)$$

has the form of the equation

$$m \ddot{\nu}(z, t) + F_0(\nu) \dot{\nu}(z, t) + \left[G_0(\nu) + EI \frac{\partial^4 \nu(z, t)}{\partial z^4} - T \frac{\partial^2 \nu(z, t)}{\partial z^2} \right] = u(z, t) \quad (6)$$

which can be seen as the generalized dynamics that govern the behavior of a second order mechanical system, with friction $F_0(\nu)$ and forces $[G_0(\nu) + EI \frac{\partial^4 \nu(z, t)}{\partial z^4} - T \frac{\partial^2 \nu(z, t)}{\partial z^2}]$ of potential functions, called *Liénard system* (Liénard (1928)).

2.3 State space representation

System (6) can be rewritten in the classical structure of a second order system in state space form, namely

$$\begin{aligned} \dot{\nu}_1 &= \nu_2 & (7) \\ \dot{\nu}_2 &= \frac{1}{m} \left[-EI \frac{\partial^4 \nu_1}{\partial z^4} + T \frac{\partial^2 \nu_1}{\partial z^2} - F_0(\nu) \nu_2 - G_0(\nu) + u(z, t) \right] \end{aligned}$$

where $\nu_1 = \nu(z, t)$, $\nu_2 = \dot{\nu}(z, t)$ and $\dot{\nu}_2 = \ddot{\nu}(z, t)$.

Now since F_0 and G_0 are linear with respect to their parameters as $F_0(\nu) = F_1^T(\nu)\theta$ and $G_0(\nu) = G_1^T(\nu)\theta$, with θ denoting the vector of parameters, the *Liénard transformation* can be applied as:

$$x_1 = \nu(z, t); x_2 = \dot{\nu}(z, t) + \frac{1}{m} \int_0^\nu F_1^T(\sigma)\theta d\sigma,$$

obtaining as a result

$$\begin{aligned} \dot{x}_1 &= x_2 - \frac{1}{m} \int_0^{x_1} F_1^T(\sigma) d\sigma \theta \\ \dot{x}_2 &= -\frac{1}{m} \left[G_1^T(x_1)\theta + EI \frac{\partial^4 x_1}{\partial z^4} - T \frac{\partial^2 x_1}{\partial z^2} - u(z, t) \right] \end{aligned} \quad (8)$$

that is:

$$\begin{aligned} \dot{x}_1 &= x_2 - \frac{1}{m} \left[cx_1 + \frac{a_1 x_1^2}{2} + \frac{a_2 x_1^3}{3} + \dots + \frac{a_\eta x_1^{\eta+1}}{\eta+1} \right] \\ \dot{x}_2 &= -\frac{1}{m} \left[EI \frac{\partial^4 x_1}{\partial z^4} - T \frac{\partial^2 x_1}{\partial z^2} + b_1 x_1 + b_2 x_1^2 + \dots + b_n x_1^n - u(z, t) \right] \end{aligned} \quad (9)$$

The PDE system (9) does not have an explicit solution. Hence, using the FDM is proposed to obtain an approximation for it, yielding for each discretization section a representation

$$\begin{aligned} \dot{x}_{1i} &= x_{2i} - \frac{1}{m} \left(cx_{1i} + \frac{a_1 x_{1i}^2}{2} + \frac{a_2 x_{1i}^3}{3} + \dots + \frac{a_\eta x_{1i}^{\eta+1}}{\eta+1} \right) \\ \dot{x}_{2i} &= -\frac{1}{m} (EI \Lambda_i - T \Upsilon_i + b_1 x_{1i} + b_2 x_{1i}^2 + \dots + b_n x_{1i}^n - u_i) \end{aligned} \quad (10)$$

where $i = 1, \dots, N$ is the index of a discretization section, and N is the number of sections,

$$\begin{aligned} \Lambda_i &= \left(\frac{x_{1(i+2)} - 4x_{1(i+1)} + 6x_{1(i)} - 4x_{1(i-1)} + x_{1(i-2)}}{(\Delta z)^4} \right), \\ \Upsilon_i &= \left(\frac{x_{1(i+1)} - 2x_{1(i)} + x_{1(i-1)}}{(\Delta z)^2} \right) \end{aligned}$$

and u_i is the discretized hydrodynamic force at section i .

Considering that displacement measurements are available over various sections, an output equation of the form:

$$y_i = x_{1i} \quad (11)$$

can be appended to (10), and terms Λ_i, Υ_i can be assumed to be functions of measurements y .

Similarly to the work presented in Fortaleza (2009), in this work the boundary conditions are $\nu(L, t) = q(t)$ (riser top), $\nu(0, t) = 0$ (riser bottom-end fixed) and $(\partial \nu / \partial z)(L, t) = (\partial \nu / \partial z)(0, t) = 0$ (rigidity condition at the fixation point)—where $q(t)$ is the function of the platform movement—for a riser with both extremities connected to fixed supports.

3. OBSERVABILITY AND IDENTIFIABILITY

On the one hand, the identification problem is closely related to an appropriate excitation condition with the choice of an identification algorithm. There are many difficulties in generalizing these conditions for nonlinear systems; however, these have been studied in the literature obtaining

conditions for some classes of nonlinear systems (Sastry and Bodson, 1989; Hammouri and Morales, 1990; Besançon et al., 1996). On the other hand, the identification problem represents the generalization of the observation problem. Occasionally physical behaviors are unknown but can be determined by employing measurements and identification techniques. A clear example is the nonlinear force identification of a marine riser; although there are models for it (e.g. Niedzwecki and Liagre (2003)), this force is generally unknown. Therefore, there is a real necessity for using nonlinear identification techniques to approximate it.

The identification problem of an unknown function can be formulated by considering the following general form of a nonlinear system:

$$\begin{aligned}\dot{x} &= f(x, u, \varphi(\theta, x, u)) \\ y &= h(x, u, \varphi(\theta, x, u))\end{aligned}\quad (12)$$

where x is the state vector, $\varphi(x, u, \theta)$ is the unknown function, θ represents an unknown state vector that characterizes the unknown function and y is the measurements' vector.

Thus, the identification problem can be reduced to reconstruct the function $\varphi(\cdot)$. This can be expanded, however, when the observation problem is associated for two reasons: (i) Suppose that the vector $x(0)$ is unknown. Then, the identification problem includes a problem of observation: one must estimate both x and $\varphi(\cdot)$. (ii) The identification topic requires an identifiability study, which is closely connected to the observability analysis, particularly when the identification techniques are based on state observers.

In Besançon et al. (2010) a state observer has been proposed for a Liénard system, because this kind of system can be put into a *state-affine* representation. Therefore, in the present work we propose the restoring force (4) that permits the transformation of system (5) into the Liénard system (10) through the transformation (8). In turn, the system (10) can be converted into a *state-affine* representation; see Besançon and Voda (2010).

Note that acceleration variables are not included in (4), because their inclusion would not allow the Liénard transformation.

Now, in order to analyze the properties of observability/identifiability of model (10)-(11) to design an observer that gives estimates of the states and the parameters, let us rewrite it as (where the section index is omitted):

$$\begin{aligned}\dot{x} &= A_o x + \Phi(y)\theta + \Phi_o(y) \\ y &= C_o x\end{aligned}\quad (13)$$

with

$$A_o = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \quad C_o = (1 \ 0), \quad y = x_1, \quad x = [x_1, x_2]^T,$$

$$\begin{aligned}\Phi(y) &= -\frac{1}{m} \begin{pmatrix} y^2 & \dots & y^{\eta+1} & \Big| & 0 & \dots & 0 \\ 2 & \dots & \eta+1 & \Big| & 0 & \dots & 0 \\ 0 & \dots & 0 & \Big| & y & \dots & y^n \end{pmatrix} \\ \theta &= [a_1, \dots, a_\eta | b_1, \dots, b_n]^T\end{aligned}\quad (14)$$

and

$$\Phi_o(y) = -\frac{1}{m} \begin{bmatrix} cy \\ E\Gamma\Lambda - T\Upsilon \end{bmatrix}.$$

Note that matrix $\Phi(y)$ can not be constructed from the system (7), i.e. without the Liénard transformation, because the polynomial $a_1x_1 + a_2x_1^2 + \dots + a_\eta x_1^\eta$ and the coefficient c are *affine* to the velocity x_2 , which is not a measurable state for estimation. Obviously if the velocity is measurable, there is not an observation problem, only an identification problem.

Now, let us consider the following persistent excitation condition in function of $y(t)$:

$$\exists T, \beta' > 0, \alpha' > 0 :$$

$$\beta' \geq \int_t^{t+T} \Psi_y^T(\tau, t) C^T C \Psi_y(\tau, t) d\tau \geq \alpha' I, \quad \forall t \geq t_0, \quad (15)$$

where Ψ_y denotes the transition matrix (as in Besançon (2007)).

Proposition 1. Assume that for any initialization of system (13), the condition (15) is satisfied; the parameter vector θ of system (13) can be then asymptotically estimated.

This result follows from a convergence condition for Kalman observers designed for *state-affine* systems (Hammouri and Morales (1990) and Besançon et al. (1996)).

Note that as demonstrated in Besançon et al. (2006) this condition is satisfied as soon as separate conditions for estimation of the state, on the one hand, and estimation of parameters, on the other hand, hold true. Similarly to the analysis of Besançon et al. (2010) and Besançon and Voda (2010), the condition for the estimation of the state is here obviously satisfied as well (A_o, C_o observable), and consequently satisfying (15) reduces to satisfying a condition for the parameter estimation.

Following Besançon et al. (2006) and Zhang (2002) such a condition can be expressed as

Υ solution of $\dot{\Upsilon} = (A_o - KC_o)\Upsilon + \Phi(y)$ for any K such that $A_o - KC_o$ is stable, if satisfies

$$\beta I \geq \int_t^{t+T} \Upsilon^T(\tau) C_o^T C_o \Upsilon(\tau) d\tau \geq \alpha I. \quad (16)$$

In that case, one can design an observer for estimation of both θ and x for instance as proposed in Zhang (2002) as

$$\begin{aligned}\dot{\hat{x}} &= A\hat{x} + \Phi(y, u)\hat{\theta} + (K + \Lambda\Gamma\Lambda^T C^T)(y - C\hat{x}) \\ \dot{\Lambda} &= (A - KC)\Lambda + \Phi(y, u) \\ \dot{\hat{\theta}} &= \Gamma\Lambda^T C^T (y - C\hat{x}).\end{aligned}\quad (17)$$

Owing to the existence of periodic solutions in Liénard systems (see e.g. Abd-Elrady et al. (2004)), one can expect that (16) will indeed be satisfied for system (13).

4. SIMULATION RESULTS

In order to program the simulator—the dynamic behavior of the riser—based on model (1), the physical parameters

listed in Table 1 are considered. The order of the spatial discretization is $n = 50$ sections, and the initial conditions are assumed to be $[\nu^{(i)}, \dot{\nu}^{(i)}]^T = [0, 0]^T$. In Fig. 2, the

Table 1. Physical parameters of the riser simulator

Parameters	Units	Values
Diameter	[m]	1.4
Riser Length	[m]	873
Mass per unit length	[kg/m]	912
Linear viscous drag coefficient	[Ns/m]	120
Duffing coefficient	[N/m ³]	8000
Van der Pol coefficient	[N/m ³]	5000
Tension	[N]	7×10^6
Bending stiffness	[Nm ²]	10^7
Inertia coefficient	-	1.05
Drag coefficient	-	1.2
Water Density	[Kg/m ³]	1,025
Current velocity	[m/s]	1.2

response of the riser when the platform oscillates periodically can be appreciated. In this case, the movement of the platform is given by the function $q(t) = 2 \sin 0.05t \times \sin 0.1t$ [m]. In order to evaluate the proposition of some restoring

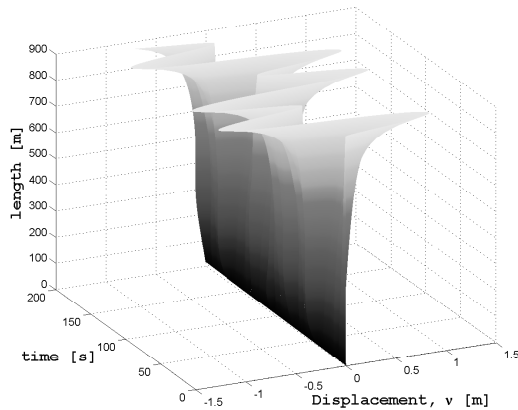


Fig. 2. Response of the riser with periodic movements of the platform.

force (4) to approximate a real unknown force, an observer with structure (17) was designed considering the following factors: (i) the unknown 'real force' was modeled by (3), (ii) the orders in the polynomials of the restoring force were chosen $\eta = n = 2$, and (iii) given the order of the polynomials, the coefficients to be estimated are a_1 , a_2 , b_1 y b_2 . In Fig. 3, the parameter estimation is exposed. Note that estimations converge to oscillating values since the estimated coefficients are not the coefficients employed to simulate the 'real force' in this test, i.e., model (5) with (3). Nevertheless, a mean ($\bar{\cdot}$) of the sustained oscillations can be computed once these have converged. Each mean can be employed to parameterize the unknown restoring force. In Fig. (4), both the estimated and 'real' simulated restoring force at some section i are shown. To obtain a better approximation, the order of the polynomial should be modified. For most structural mechanic systems, odd polynomials are a suitable choice (e.g. third order models without second order terms). In general, not all the terms

are significant; and a selection can be made by simply checking the relative influence of each term by a preliminary identification and by retaining the most important ones (Ceravolo et al. (2013)).

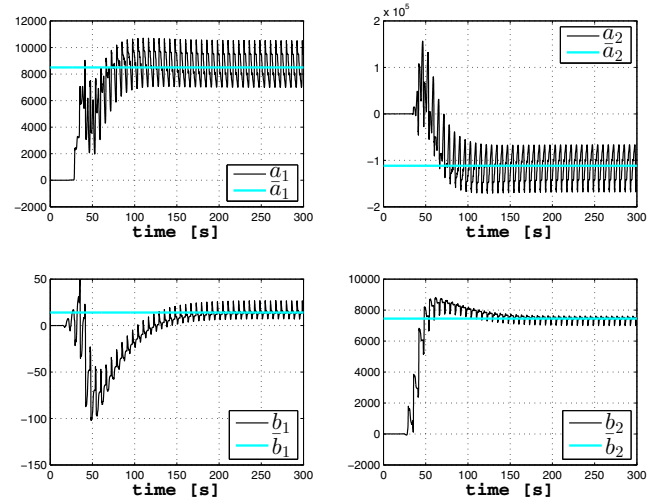


Fig. 3. Estimation of a_1 , a_2 , b_1 and b_2 .

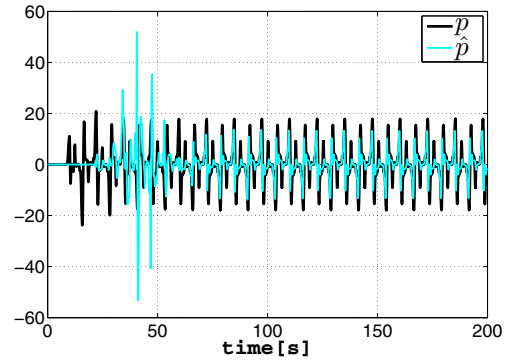


Fig. 4. Estimation of the restoring force.

5. CONCLUSION

This paper introduced a nonlinear restoring model for vertical risers based on polynomial functions. Such a restoring model permits the transformation of the riser model into a Liénard representation, which is suitable for the conception of observers that identify structural parameters or the riser restoring force when its physical model is not available. Although the identification technique presented here has been employed in a specific application, it can be extended for being applied in other mechanical systems; but with the caveat that a polynomial model should be used for small nonlinearities or as a "first trial" when a pertinent parametric model is unknown. Finally, in future works, the use of the presented approach would be interesting in developing schemes for the detection of structural damages.

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