# Benjamin Fasquelle

e-mail: Benjamin.Fasquelle@ls2n.fr

Ecole Centrale de Nantes, Centre National de la Recherche Scientifique, Laboratoire des Sciences du Numérique de Nantes (LS2N), UMR CNRS 6004, 1 rue de la Noë, Nantes 44321, France

# Parag Khanna

Ecole Centrale de Nantes,
Centre National de la Recherche Scientifique,
Laboratoire des Sciences du Numérique de
Nantes (LS2N),
UMR CNRS 6004, 1 rue de la Noë,
Nantes 44321, France
e-mail: parag.khanna@ls2n.fr

# **Christine Chevallereau**

Ecole Centrale de Nantes,
Centre National de la Recherche Scientifique,
Laboratoire des Sciences du Numérique de
Nantes (LS2N),
UMR CNRS 6004, 1 rue de la Noë,
Nantes 44321, France
e-mail: Christine.Chevallereau@ls2n.fr

# **Damien Chablat**

Ecole Centrale de Nantes, Centre National de la Recherche Scientifique, Laboratoire des Sciences du Numérique de Nantes (LS2N),

UMR CNRS 6004, 1 rue de la Noë, Nantes 44321, France e-mail: damien.chablat@ls2n.fr

## **Denis Creusot**

Ecole Centrale de Nantes, Centre National de la Recherche Scientifique, Laboratoire des Sciences du Numérique de Nantes (LS2N),

UMR CNRS 6004, 1 rue de la Noë, Nantes 44321, France e-mail: denis.creusot@ls2n.fr

# Stéphane Jolivet

Ecole Centrale de Nantes, Centre National de la Recherche Scientifique, Laboratoire des Sciences du Numérique de Nantes (LS2N), UMR CNRS 6004, 1 rue de la Noë,

JMR CNRS 6004, 1 rue de la Noë, Nantes 44321, France e-mail: Stephane.Jolivet@ls2n.fr

# Philippe Lemoine

Ecole Centrale de Nantes,
Centre National de la Recherche Scientifique,
Laboratoire des Sciences du Numérique de
Nantes (LS2N),
UMR CNRS 6004, 1 rue de la Noë,
Nantes 44321, France
e-mail: Philippe.Lemoine@ls2n.fr

# Identification and Control of a 3-X Cable-Driven Manipulator Inspired From the Bird's Neck

This paper is devoted to the control and identification of a manipulator with three anti-parallelogram joints in series, referred to as X-joints. Each X-joint is a tensegrity one-degree-of-freedom mechanism antagonistically actuated with cables and springs in parallel. As compared to manipulators built with simple revolute joints in series, manipulators with tensegrity X-joint offer a number of advantages, such as an intrinsic stability, variable stiffness, and lower inertia. This design was inspired by the musculosleketon architecture of the bird's neck, which is known to be very dextrous. A test-bed prototype is presented and used to test computed torque control laws. Friction and cable elasticity are modeled and identified. Their effect on the performance of control laws is analyzed. It is shown that in the context of antagonistic actuation and lightweight design, friction plays a leading role and the significance of modeling cable elasticity is discussed. [DOI: 10.1115/1.4051521]

Keywords: control, identification, bio-inspired design, tensegrity mechanism

<sup>&</sup>lt;sup>1</sup>Corresponding author.

Contributed by the Mechanisms and Robotics Committee of ASME for publication in the JOURNAL OF MECHANISMS AND ROBOTICS. Manuscript received February 17, 2021; final manuscript received June 9, 2021; published online July 6, 2021. Assoc. Editor: Chin-Hsing Kuo.

# Philippe Wenger

Ecole Centrale de Nantes,
Centre National de la Recherche Scientifique,
Laboratoire des Sciences du Numérique de
Nantes (LS2N),
UMR CNRS 6004, 1 rue de la Noë,
Nantes 44321, France
e-mail: Philippe.Wenger@ls2n.fr

### 1 Introduction

Bio-inspired design has drawn the interest of many researchers and is an interesting avenue for designing light-weight manipulators. Such manipulators are desired in a number of applications, in particular when safe interactions or high dynamics are required. Cable-driven manipulators, which mimic the muscles-and-tendons actuation system met in animals, are good candidates for lightweight design, since the motors can be fixed on the ground. Two main families are considered in the literature, namely, continuum manipulators and discrete manipulators. Inspired from the elephant trunk or the octopus tentacle, the former has a flexible structure and, consequently, infinitely many degrees-of-freedom. They find applications in highly cluttered environments and their accurate modeling is a real challenge [1]. The latter consists of a discrete number of joints, as in the musculoskeletal system of vertebrates. They can be more easily modeled and are still interesting solutions for designing dextrous manipulators [2,3]. Applications include environmentally interactive robots [4], robotic hands [5], exoskeletons [6], and other medical devices [7].

The general frame of this work is the design and control of a robotic manipulator inspired by the bird's neck. Birds use their neck like an arm for every-day-life tasks. Cleaning and feeding necessitate dexterity while high force transmission and accelerations are required when a woodpecker hits a tree trunk. We have built a preliminary manipulator model to mimic the neck of a woodpecker, which moves mainly in a vertical plane [8,9]. Bird necks have a variable number of vertebrae, ranging from 10 in parrots to 26 in swans [10]. This preliminary model uses three joints. This choice was not only motivated to start with a simple design: from a biological point of view, indeed, the bird's neck can be regarded as a serial assembly of three main sets of vertebrae, each set having its own function [10]. The prototype consists of the serial assembly of three tensegrity joints remotely actuated with cables. A tensegrity structure is an assembly of compressive elements (bars) and tensile elements (cables, springs) held together in equilibrium [11,12]. Tensegrity is known in architecture and art for more than a century [13] and is suitable for aerospace applications [14,15] and for modeling living organisms [16]. Tensegrity mechanisms have been more recently studied for their promising properties in robotics such as low inertia, natural compliance, and deployability [17–19]. The tensegrity joints used in our neck model are anti-parallelogram joints inspired from the Snelson's X-shape mechanisms [20], referred to as *X-joints* in this paper. As compared with simple revolute joints, X-joints do not have a fixed rotation axis, a common feature in biological joints [9,21]. Moreover, they have a number of advantages such as larger rotation ranges [4] and larger variations of stiffness [22].

Cable-driven manipulators raise a number of difficulties, such as the choice of suitable routing strategies [2,3,23] and accounting for cable flexibility [24–26]. Moreover, the control of cable-driven manipulators is also a challenging task [27,28]. In Ref. [7], a control strategy to compensate friction was proposed but it relied on a measurement of cable tension, which adds complexity and is not available in our prototype. In other papers such as in Ref. [29], force sensors are used for the identification and a compensation of the identified model is then integrated by control. In the proposed paper, the cable-driven robot used is not a cable-driven

parallel robot such as in Ref. [26] nor a tendon-based actuation system with tendons sliding in curved pathways such as in Ref. [29]. Accordingly, the effects of cable elongation modeling differ. Moreover, our identification method and control strategy do not use any tension sensors. This paper is devoted to the identification and control of a 3-X manipulator remotely actuated with cables. Two main sets of parameters are identified: geared motor frictions and cable stiffnesses. The identification has a variety of objectives such as to better understand our system, to build a simulator that is closer to the experiments or to test several control strategies. Periodic identification campaigns can also be conducted for the purpose of monitoring cable wear. Model-based control laws, which use the dynamic model of the complete system, are implemented and tested on a testbed prototype. We want to explore the impact of the identified friction and cable stiffness models on the performance of control law.

Section 2 describes the experimental setup. Section 3 presents the cable transmission and dynamic models. Calculation of the manipulator joint configuration from the actuator encoders is presented in Sec. 4. Friction and cable elongation models as well as their associated identification models are discussed in Sec. 5. Section 6 proposes computed torque control laws and test results are discussed in Sec. 7. Last section concludes this paper.

## 2 Experimental Setup

**2.1 Manipulator Architecture.** The experimental set up is a series assembly of three cable-driven X-joints. Each joint is equipped with two lateral springs of equal stiffness and assumed mass-less (Fig. 1). Springs have been often used in tensegrity mechanisms [17–19]. The role of springs is to assure a minimal stiffness. Spring stiffnesses used in our prototype were chosen to have a stable equilibrium at rest, and their rest length  $l_0$  was chosen so that the springs remain always in tension. Finally, we came up with a set of commercial springs with the following features: two springs of stiffness  $300 \text{ N m}^{-1}$  arranged in parallel on each side of the first X-joint (defining an equivalent  $600 \text{ N m}^{-1}$  stiffness spring on each side) and two springs of stiffness  $100 \text{ N m}^{-1}$  arranged in parallel on each side of the second and third X-joint (defining an equivalent  $200 \text{ N m}^{-1}$  stiffness spring on each side); rest length  $l_0 = 0.046$  for all springs.

To avoid self-collisions, the crossing bars have been assembled in two parallel planes and the top and base bars are present in each plane. This arrangement also reduces deformation out of the vertical plane. In the original Snelson X-shape mechanism, the top and bottom sides are cables or springs. Here, only one-degreeof-freedom is desired. To avoid any undesired rotation of the diagonal bars, the top and bottom cables have been replaced with rigid bars. The resulting mechanism is still a tensegrity one, but it is of class 2 (for the first and third joint) and 3 (for the second joint), instead of class 1 [13]. The top bar of the first joint (resp. second joint) is the same as the bottom bar of the second joint (resp. third joint). The configuration of each joint i is defined by the orientation angle  $\alpha_i$  of its top bar with respect to its base bar. All revolute joints connecting the links are assumed perfect. The two diagonal bars have the same length ( $L = 100 \,\mathrm{mm}$ ) while the top and base bars have the same length  $b = 50 \,\mathrm{mm}$ .

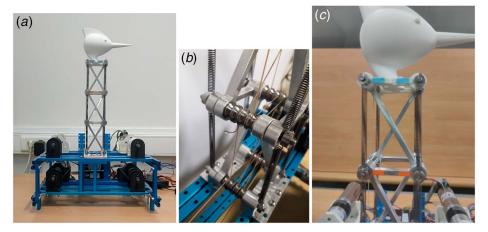


Fig. 1 (a) Our prototype with its three X-joints and four motors, (b) pulley arrangement, and (c) colored markers on base and top bars of the third mechanism

2.2 Actuation. The three joints are driven remotely with cables controlled by winches mounted on rotary motors. Ball bearing-mounted loose pulleys are used to guide the cables from the winches to the X-joints. All pulleys are mounted on the axes linking the rigid bars and arranged in four rows. Thus, each cable has its own row and does not cross each other, as shown in Fig. 1. The number of active cables needs to be determined properly. We want to control the orientation angle of the three X-joints. In theory, a minimum of three active cables is required. In practice, at least one more cable is required to produce a sufficiently large wrench-feasible workspace [30,23]. The maximum number is six, a scheme where each X-joint would be antagonistically actuated with two cables, one on each side. We propose to use four motors. This choice is a compromise that allows controlling the three joints with a reasonable number of motors. Several cable routing strategies can be considered, including strut-routed (i.e., passing along rigid bars [31]) and side-routed (i.e., running on sides through springs [4,23]). We propose to use a hybrid strut-routed/side-routed strategy. One long cable is run in a siderouted way through the left sides of the three X-joints and is attached to the last X-joint (see Fig. 2). This cable actuates the three X-joints simultaneously. On the right sides, three cables are run through the X-joints and attached to the right tip of the top bar of the driven X-joint. If there were no pulleys, each of those three cables would move solely one X-joint. As it will be shown

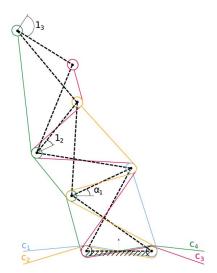


Fig. 2 Actuation scheme of the 3-X cable driven manipulator prototype

in next section, however, the presence of pulleys induces slight couplings. In the woodpecker neck, the long, left side-routed cable corresponds to a powerful muscle found all along the front of the neck. This muscle allows the neck to quickly propel its head to hit tree trunks [8]. The other cables correspond to other muscles present to adjust the movements. The path of each cable has been adjusted to avoid any loss of contact between pulleys and cables. In particular, the green cable wraps around the two intermediate pulleys by making at least one turn. Each cable is attached to a drum linked to a motor shaft. The first motor pulls the side-routed green cable, which wraps around the two left-sided intermediate pulleys to ensure contact. Each of the other three motors pulls one X-joint on its right. The yellow and red cables that drive the second and third joint, respectively, are strut routed. All pulleys have a radius of  $r = 4.6 \,\mathrm{mm}$  and the four drums have a radius of  $r_d = 20 \,\mathrm{mm}$ . The four motors are MAXON brushless gearmotors (ECMAX40L/PM42) with a gear ratio of R = 25. Four ESCON 50/5 controllers ensure velocity and torque control of the motors. The frequency of the control law is 500 Hz. The motors interact with a Beaglebone (BB) blue microcontroller, on which the control law is programmed.

# 3 Cable Length and Dynamic Models

This section is devoted to the derivation of the cable length and dynamic models of the manipulator. The input variables of our system are the forces applied by the motors on the cables. Our goal is to control the orientation angles  $\alpha_i$  of each X-joint. The X-joints are numbered from bottom to top. Since the only measured variables are the motor angular positions  $\theta_i$ ,  $\alpha_i$  must be computed from  $\theta_i$ . To this end, the relationship between the cable lengths and the orientation angles of the X-joints is first established.

**3.1 Cable Length Model.** Pulleys used to guide the cables from the winches to the X-joints affect the calculation of the unrolled cable lengths required for desired orientation angles  $\alpha_i$  by adding some coupling between the X-joints. In Ref. [9], a simplified cable length model was used by neglecting the radii of all pulleys, which induced some errors in the model. We now derive the expression of each unrolled cable length as a function of the orientation angle  $\alpha_i$  by taking into account the pulleys properly.

Let  $l_l(\alpha_i)$  and  $l_r(\alpha_i)$  be the left and right side lengths of each X-joint, respectively, which correspond to the spring lengths. Their expression is [23]

$$l_l(\alpha_i) = \sqrt{L^2 - b^2 \cos^2\left(\frac{\alpha_i}{2}\right)} - b \sin\left(\frac{\alpha_i}{2}\right) \tag{1}$$

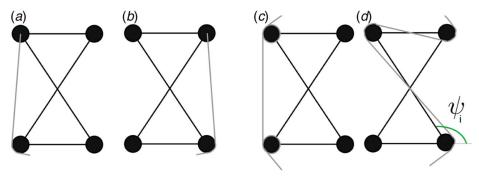


Fig. 3 The four cases met for cable variable lengths calculation, referred to as (a)–(d) from left to right. The variable part of the cable length as a function of  $\alpha_i$  is defined by Eqs. (3)–(6) for cases (a) to (d), respectively.

$$l_r(\alpha_i) = \sqrt{L^2 - b^2 \cos^2\left(\frac{\alpha_i}{2}\right)} + b \sin\left(\frac{\alpha_i}{2}\right)$$
 (2)

The unrolled length of each cable can be expressed as the sum of a variable length that depends on  $\alpha_i$  and on a constant length that needs not be determined. The variable length calculation depends on the cable way along each X-joint. Pulleys used to guide cables must be taken into account and add some complexity in the calculations. A kinematic model of the cables considering pulleys was derived in Ref. [32] for a cable-driven parallel manipulator. For the manipulator at hand, four different cases must be considered, as shown in Fig. 3. For the first two cases shown on the left of Fig. 3, referred to as (a) and (b), the left and right variable length  $l_i(\alpha_i)$  can be written as, respectively (see Ref. [33] for more details):

$$l_i(\alpha_i) = \sqrt{l_l(\alpha_i)^2 - r^2} + r \left( -\arctan\left(\frac{\sqrt{l_l(\alpha_i)^2 - r^2}}{r}\right) - \frac{\alpha_i}{2} \right)$$
 (3)

$$l_i(\alpha_i) = \sqrt{l_r(\alpha_i)^2 - r^2} + r \left(\frac{\alpha_i}{2} - \arctan\left(\frac{\sqrt{l_r(\alpha_i)^2 - r^2}}{r}\right)\right)$$
(4)

For the third case, referred to as (c) in Fig. 3, the variable length is [33]

$$l_i(\alpha_i) = l_l(\alpha_i) - r\alpha_i \tag{5}$$

Finally, the variable length associated with the fourth case (Fig. 3, right), referred to as (d), corresponds to the strut routing scheme and can be expressed as [33]

$$l_i(\alpha_i) = r(2\psi(\alpha_i) - \alpha_i) \tag{6}$$

where  $\psi(\alpha_i)$  is the angle between the diagonal bar and the base bar as shown in Fig. 3, right. It can be expressed as follows [34]:

$$\psi_i = 2\arctan\left(\frac{-2L\sin(\alpha_i) - \sqrt{(2L\sin(\alpha_i))^2 + (2l(\cos(\alpha_i) + 1))^2 - (2b(\cos(\alpha_i) + 1))^2}}{(2b - 2L)(\cos(\alpha_i) + 1)}\right)$$
(7)

To obtain the total variable length associated with each cable, the appropriate cable case (a), (b), (c), or (d) along each X-joint must be first identified. Then, the associated variable lengths must be added together.

The blue cable  $c_1$  is attached to the first X-joint (Fig. 2) and its way simply follows case (b) along this sole joint. Thus, its variable length  $l_{c1}$  can be directly computed from Eq. (3) by

$$l_{c1} = \sqrt{l_l(\alpha_1)^2 - r^2} + r \left(\frac{\alpha_1}{2} - \arctan\left(\frac{\sqrt{l_l(\alpha_1)^2 - r^2}}{r}\right)\right)$$
(8)

The yellow cable  $c_2$  is attached to the second X-joint (see Fig. 2) and follows a way corresponding to case (d) along the first X-joint (Fig. 3, right), and to case (b) along the second X-joint (Fig. 3, left). Its variable length  $l_{c2}$  can thus be written as

$$l_{c2} = \sum_{i=1}^{2} l_i(\alpha_i) \tag{9}$$

with

$$l_1(\alpha_1) = r(2\psi_1 - \alpha_1) \tag{10}$$

$$l_2(\alpha_2) = \sqrt{l_r(\alpha_2)^2 - r^2} + r\left(\frac{\alpha_2}{2} - \arctan\left(\frac{\sqrt{l_r(\alpha_2)^2 - r^2}}{r}\right)\right)$$
(11)

The red cable  $c_3$  follows a way similar to the yellow cable, with the difference that it has to go up to the third X-joint (see Fig. 2). Its variable length  $l_{c_3}$  can be written as

$$l_{c3} = \sum_{i=1}^{3} l_i(\alpha_i)$$
 (12)

where

$$l_1(\alpha_1) = r(2\psi_1 - \alpha_1) \tag{13}$$

$$l_2(\alpha_2) = r(2\psi_2 - \alpha_2) \tag{14}$$

$$l_3(\alpha_3) = \sqrt{l_r(\alpha_3)^2 - r^2} + r\left(\frac{\alpha_3}{2} - \arctan\left(\frac{\sqrt{l_r(\alpha_3)^2 - r^2}}{r}\right)\right)$$
(15)

Finally, the green cable running along the left side of the manipulator, referred to as  $c_4$  (see Fig. 2) is attached to the last X-joint and follows a way corresponding to case (c) along the first two X-joints (Fig. 3, middle) and to case (a) along the

last X-joint (Fig. 3, left). Accordingly, the variable length  $l_{c4}$  of this cable can be written as

 $l_{c4} = \sum_{i=1}^{3} l_i(\alpha_i) \tag{16}$ 

where

$$l_1(\alpha_1) = l_l(\alpha_1) - r\alpha_1 \tag{17}$$

$$l_2(\alpha_2) = l_l(\alpha_2) - r\alpha_2 \tag{18}$$

$$l_3(\alpha_3) = \sqrt{l_l(\alpha_3)^2 - r^2} + r \left( \arctan\left(-\frac{\sqrt{l_l(\alpha_3)^2 - r^2}}{r}\right) - \frac{\alpha_3}{2} \right)$$
 (19)

Let us set  $\theta_i = 0$ , i = 1,..., 4 and  $\mathbf{I}_c = \mathbf{I}_{c0}$  when  $\alpha_i = 0$ , i = 1,..., 3 (the manipulator is at rest, i.e., straight right).

Assuming inextensible cables, the motor positions can be expressed as functions of the cable lengths by the following vector equation:

$$\theta = -\frac{R}{r_d}(\mathbf{l}_c - \mathbf{l}_{c0}) \tag{20}$$

where  $\theta = [\theta_1, \theta_2, \theta_3, \theta_4]^T$ ,  $\mathbf{l}_c = [l_{c1}, l_{c2}, l_{c3}, l_{c4}]^T$ , and  $\mathbf{l}_{c0}$  is the value of  $\mathbf{l}_c$  at rest. We assume that  $\theta_i > 0$  when motor i winds up cable  $c_i$ ; hence, the minus sign in the above expression. As a reminder,  $r_d$  is the drum radius and R is the reduction ratio.

As a result, the motor velocities  $\dot{\theta}$  can be expressed as functions of the X-joint velocities  $\dot{\alpha}$  as

$$\dot{\theta} = \frac{R}{r_d} \mathbf{Z}^T(\alpha) \dot{\alpha} \tag{21}$$

where **Z** is the cable transmission  $3 \times 4$  matrix, whose elements are  $Z_{ij}$ :

$$\mathbf{Z}_{ij} = -\frac{\partial l_c(j)}{\partial \alpha_i} \tag{22}$$

This matrix can be decomposed into the form:

$$\mathbf{Z} = (\mathbf{Z}_1 \ \mathbf{Z}_2) \tag{23}$$

where  $\mathbf{Z}_1$  is a  $3\times3$  triangular matrix associated with the blue, yellow, and red cables pulling on the right, and  $\mathbf{Z}_2$  is a vector of size 3 associated with the green cable pulling on the left. Thus, the diagonal entries of  $\mathbf{Z}_1$  are of constant sign opposite to the sign of the components of  $\mathbf{Z}_2$  and  $\mathbf{Z}_1$  is invertible. Moreover, the non-diagonal entries of  $\mathbf{Z}_1$ , which are due to the pulleys, are small as compared to the diagonal entries because the pulleys' radii are small as compared to the X-joint size. This decomposition will be used in Sec. 6 for control.

**3.2 Dynamic Model.** Let **M** be the inertia matrix accounting for the manipulator inertia  $\mathbf{M}_m$  [35] and for the geared motors inertia:

$$\mathbf{M}(\alpha) = \mathbf{M}_{m}(\alpha) + I_{a} \left(\frac{R}{r_{d}}\right)^{2} \mathbf{Z}(\alpha) \mathbf{Z}(\alpha)^{T}$$
(24)

where  $I_a$  is the inertia of the geared motors. The inertia matrix  $\mathbf{M}(\alpha)$  is deduced from the kinetic energy. All geared motors are identical and their total contribution to the kinetic energy is  $\frac{I_a}{2}\dot{\theta}^T\dot{\theta}$ . It can also

be written  $\frac{I_a}{2} \left( \frac{R}{r_d} \right)^2 \dot{\alpha}^T \mathbf{Z}(\alpha) \mathbf{Z}(\alpha)^T \dot{\alpha}$  using the relationship between motors and X-joints rotation angles given in Eq. (21). Neglecting friction and cable elasticity, a first dynamic model can be derived

as follows [9]:

$$\mathbf{M}(\alpha)\ddot{\alpha} + \mathbf{c}(\alpha, \dot{\alpha}) + \mathbf{g}(\alpha) = \mathbf{Z}(\alpha)\mathbf{f}$$
 (25)

where **g** is the gradient of the manipulator potential energy with respect to the generalized coordinates, its includes the gravity and spring forces,  $\mathbf{f} = [f_1, f_2, f_3, f_4]^T$  is the vector of motor forces transmitted by the cables and **c** is the vector of Coriolis effects affecting the manipulator,  $\mathbf{c}_m$ , and the geared motors:

$$\mathbf{c} = \mathbf{c}_m(\alpha) + I_a \left(\frac{R}{r_d}\right)^2 \dot{\mathbf{Z}}(\alpha, \dot{\alpha}) \mathbf{Z}(\alpha)^T$$
 (26)

where the second term of LHS (left-hand side) of Eq. (26) comes from the contributions of the geared motors to the kinetic energy. The control law tested in Ref. [9] was based on this dynamic model but the cable transmission matrix Z was calculated by neglecting the effects of pulleys. With this control law, large oscillations in the applied forces were observed and values of  $\alpha$  computed by the control were significantly different from the observed ones. In order to obtain a better control, an improved dynamic model is proposed thereafter, including pulley effects, motor friction, and cable elasticity. Friction and cable elasticity must be identified. All other parameters appearing in the dynamic model are available from the CAD model of our prototype and from motor data-sheets. Since the manipulator is intrinsically light and because of the reduction gear, the contribution of  $\mathbf{M}_m$  in Eq. (24) is low as compared to the contribution inertia of the geared motors. On the other hand, this latter cannot be assumed constant as it is generally the case in manipulators that are not driven with cables. Indeed, matrix **Z** depends on the cable routing and on  $\alpha$ . The inertia matrix **M** is therefore variable, and it is important to take it into account in our models and in the control.

**3.3 Dynamic Model With Friction and Elastic Cables.** All cables are now considered elastic. Assuming  $\theta_i = 0$ , i = 1,...,4 when  $\alpha_i = 0$ , i = 1,...,3 at rest (i.e., with zero forces applied and, thus, with zero cable elongations), the vector of cable elastic elongations  $\mathbf{x}_c = [x_{c1}, x_{c2}, x_{c3}, x_{c4}]^T$  is defined as

$$\mathbf{x}_c = \mathbf{l}_c - \mathbf{l}_{c0} - \frac{r_d}{R} \theta \tag{27}$$

In other words, the elongation of each cable is defined as the difference between its theoretical length as calculated in Sec. 3.1, and the cable length wrapped around the associated winch.

Upon differentiating Eq. (27) with respect to time, we obtain

$$\mathbf{Z}(\alpha)^T \dot{\alpha} = \frac{r_d}{R} \dot{\theta} + \dot{\mathbf{x}}_c \tag{28}$$

Let  $\mathbf{t}_c = [t_{c1}, \ t_{c2}, \ t_{c3}, \ t_{c4}]^T$  define the vector of cable tensions and let  $\mathbf{f}_{friction}$  be the vector of friction forces. Friction mostly occurs in the geared motors. Modeling friction in pulleys can be of interest to improve measurement accuracy of cable force sensors [36]. In this paper, no cable force sensors are used. We thus concentrate on modeling friction due the geared motors. Writing the dynamic model for the system made of the geared motors and of the cables, provides the following motion equation of  $\theta$ :

$$I_a \ddot{\theta} + \frac{r_d}{R} \mathbf{t}_c + \frac{r_d}{R} \mathbf{f}_{friction} = \frac{r_d}{R} \mathbf{f}$$
 (29)

The dynamic model written for the manipulator and the cables gives the motion equation of  $\alpha$ :

$$\mathbf{M}_{m}(\alpha)\ddot{\alpha} + \mathbf{c}_{m}(\alpha, \dot{\alpha}) + \mathbf{g}(\alpha) = \mathbf{Z}(\alpha)\mathbf{t}_{c}$$
 (30)

Upon differentiating Eq. (28) with respect to time and substituting  $\ddot{\theta}$  into Eq. (29), cable tension can be expressed as follows:

$$\mathbf{t}_{c} = \mathbf{f} - \mathbf{f}_{friction} - I_{a} \left(\frac{R}{r_{d}}\right)^{2} (\mathbf{Z}(\alpha)^{T} \ddot{\alpha} + \dot{\mathbf{Z}}(\alpha)^{T} \dot{\alpha} - \ddot{\mathbf{x}}_{c})$$
(31)

Substituting the above expression into Eq. (30) gives the dynamic model of the full system with friction and cable elasticity:

$$\mathbf{M}(\alpha)\ddot{\alpha} + \mathbf{c}(\alpha, \dot{\alpha}) + \mathbf{g}(\alpha) = \mathbf{Z}(\alpha) \left( \mathbf{f} - \mathbf{f}_{friction} + I_a \left( \frac{R}{r_d} \right)^2 \ddot{\mathbf{x}}_c \right)$$
(32)

Friction and cable elasticity must be then identified.

#### 4 Estimation of $\alpha : \hat{\alpha}$

Our prototype is not equipped with encoders to measure  $\alpha$  online.  $\hat{\alpha}$  can be reconstructed from  $\theta$  (measured by the motor encoders) through cable lengths with the cable models derived in Sec. 3.1. On the other hand, cable elasticity must be taken into account. Consequently, cable stiffness must be identified. To this end, the elongation of the cables are calculated from external measurements of  $\alpha$  using video recording during the identification experiments.

**4.1** Calculation of  $\hat{\alpha}$  From  $\theta$ . Two methods are considered to compute  $\hat{\alpha}$  from the measured values of  $\theta$ . We assume in this section that the cable elongation  $\mathbf{x}_c$  and its derivative are known from the identified cable stiffness and from the cable tensions. The proposed methods can be also used with inelastic cables by setting  $\mathbf{x}_c = 0$ .

In the first method, the minimal number of measurement data of  $\theta$ , namely, 3, is used. The estimation  $\hat{\alpha}$  then relies on the model of cable elasticity and this method can be used to evaluate the quality of this model.

The second method uses all the four available measurement data of  $\theta$ . The estimation  $\hat{\alpha}$  is defined as a compromise between these four information data. Accordingly, the estimation reliability is less dependent on the quality of the model of cable elasticity.

4.1.1 Calculation From Three Components of  $\theta$ . The first method uses only three components of  $\theta$ . The idea is to calculate the X-joint orientations  $\hat{\alpha}_1$ ,  $\hat{\alpha}_2$  and  $\hat{\alpha}_3$  recursively as functions of the lengths of right cables (i.e., the blue, yellow, and red cables used to actuate the first, second, and third X-joint). The X-joint orientation  $\hat{\alpha}_1$  is first computed from the length  $l_{c1}$  of the blue cable (Fig. 2) using Eqs. (8) and (27). Knowing  $\hat{\alpha}_1$ ,  $\hat{\alpha}_2$  is then computed from the length  $l_{c2}$  of the yellow cable using Eqs. (9) and (27).

Knowing  $\hat{\alpha}_1$  and  $\hat{\alpha}_2$ ,  $\hat{\alpha}_3$  is finally computed from the length  $l_{c3}$  of the red cable with Eqs. (12) and (27). The following triangular system is obtained:

$$\begin{cases}
\hat{\alpha}_{1} = f^{1}(\theta_{1}, x_{c1}) \\
\hat{\alpha}_{2} = f^{2}(\theta_{1}, \theta_{2}, x_{c1}, x_{c2}) \\
\hat{\alpha}_{3} = f^{3}(\theta_{1}, \theta_{2}, \theta_{3}, x_{c1}, x_{c2}, x_{c3})
\end{cases}$$
(33)

No closed-form expressions could be found for this system. However, owing to the monotony of the cable length functions used to obtain the above system, we know that  $\hat{\alpha}_1$ ,  $\hat{\alpha}_2$ , and  $\hat{\alpha}_3$  are defined uniquely. We have implemented a simple, fast procedure to obtain approximated values of  $\hat{\alpha}_i$  by linear interpolation from a pre-calculated set of solutions. The number of pre-calculated solutions is sufficiently large so that the error of  $\hat{\alpha}_i$  is always lower than the resolution of the motor encoders.

4.1.2 Calculation From All Components of  $\theta$ . The second method to obtain  $\hat{\alpha}$  from  $\theta$  uses all components of  $\theta$ . Knowing that the cables are elastic, the idea is to use all the measurements to have a more realistic computation.

Equation (28) can be rewritten as

$$\mathbf{Z}(\hat{\alpha}(t))^{T}(\hat{\alpha}(t+\Delta t) - \hat{\alpha}(t)) = \frac{r_d}{R}(\theta(t+\Delta t) - \theta(t)) + (\mathbf{x}_c(t+\Delta t) - \mathbf{x}_c(t))$$
(34)

where  $\Delta t$  is a sufficiently small time-step. The Moore–Penrose pseudo-inverse  $\mathbf{Z}^{T+}$  allows computing the variation in  $\hat{\alpha}$  that minimizes the error. Accordingly,  $\hat{\alpha}$  can be computed as follows:

$$\hat{\alpha}(t + \Delta t) = \hat{\alpha}(t) + \mathbf{Z}(\hat{\alpha}(t))^{T+}$$

$$\left[\frac{r_d}{R}(\theta(t + \Delta t) - \theta(t)) + (\mathbf{x}_c(t + \Delta t) - \mathbf{x}_c(t))\right]$$
(35)

4.2 Measuring  $\alpha$  With Image Processing:  $\alpha_{IP}$ . Identification of cable stiffness required knowledge of cable elongation and, in turn, of both  $\theta$  and  $\alpha$ . The former is measured from the motor encoders and the latter is measured by video. We use a camera with an image size of 1280×720 pixels and 30 frames per second. During experiments, X-joints displacements are recorded. To this end, several colored markers are placed on top and base bars of the three X-joints (see Fig. 1). Then,  $\alpha_{IP}$  is computed offline with an image processing (IP) algorithm on the basis of the markers. The target area for the IP analysis is kept in the center of video to avoid any distortion. The first and second derivatives of  $\alpha_{IP}$  are computed numerically. Note that the low video frequency does not allow using  $\alpha_{IP}$  for control purposes, but it can be used for the identification experiments. The precision of the image processing is about 1 deg. This precision is acceptable in view of the errors encountered during control (see Sec. 6).

# 5 Identification of Friction and Cable Stiffness Parameters

- **5.1 Identification Model.** Friction and elastic cable models are established in this section. We then derive a system of equations linear in the parameters to identify. We use an identification method successfully applied in Refs. [37,38]. This method is detailed in Ref. [39].
- 5.1.1 Friction Equations. As written previously, friction is assumed to occur mainly in the geared motors. A dry friction model can be written as follows:

$$\mathbf{f}_{friction}(\mathbf{f}_s, \dot{\theta}) = \operatorname{diag}(\operatorname{sign}(\dot{\theta}))\mathbf{f}_s$$
 (36)

where  $\mathbf{f}_s = [f_s(1), f_s(2), f_s(3), f_s(4)]^T$  is a constant vector that must be identified,  $diag(\mathbf{u})$  is a diagonal  $(4 \times 4)$  matrix built with the components of  $\mathbf{u}$  and sign is the sign function. A model ensuring continuity can be derived as follows:

$$\mathbf{f}_{friction}(\mathbf{f}_s, \dot{\theta}) = \operatorname{diag}\left(\frac{2}{\pi}\arctan\left(c_s\dot{\theta}\right)\right)\mathbf{f}_s$$
 (37)

where  $c_s$  is a tuning parameter that allows adjusting the slope of the arc-tangent around 0. This parameter is fixed arbitrarily to  $c_s = 0.5$ .

Factor  $\frac{2}{\pi}$  ensures that  $\mathbf{f}_{friction}$  tends toward  $\mathbf{f}_{s}$  and  $-\mathbf{f}_{s}$  when  $\dot{\theta}$  has large positive and negative values, respectively.

Substituting  $\mathbf{f}_{friction}$  into Eq. (32), we obtain

$$\mathbf{Z}(\alpha)\operatorname{diag}\left(\frac{2}{\pi}\arctan\left(c_{s}\dot{\theta}\right)\right)\mathbf{f}_{s} = \mathbf{Z}(\alpha)\left(\mathbf{f} + I_{a}\left(\frac{R}{r_{d}}\right)^{2}\ddot{\mathbf{x}}_{c}\right) - \mathbf{M}(\alpha)\ddot{\alpha} - \mathbf{c}(\alpha, \dot{\alpha}) - \mathbf{g}(\alpha) \quad (38)$$

We thus obtain a system of three equations that are linear in  $\mathbf{f}_s$ . Once  $\alpha$ ,  $\theta$ , their derivatives and  $\mathbf{f}$  are known,  $\mathbf{f}_s$  can thus be identified. A model with viscous friction was tested but this

model did not provide better results, confirming that dry friction is preponderant [33].

5.1.2 Cable Stiffness Equations. During the identification experiments,  $\alpha$  is obtained from the image processing  $\alpha = \alpha_{IP}$  and  $\theta$  is measured by the motor encoders. Cable elongation  $\mathbf{x}_c$  can be computed with Eq. (27).

This calculation assumes that there is no cable elongation in the initial manipulator configuration  $\alpha_i = 0$ , i = 1, ..., 3. In practice, the cables need to be slightly in tension to avoid slack, but the elongation is small and can be neglected. The derivative  $\dot{\mathbf{x}}_c$  of  $\mathbf{x}_c$  is computed numerically.

Rather than resorting to an elastic Young modulus based model [40], we prefer to write a more general cable model that makes it possible to identify a set of parameters for each cable:

$$\mathbf{t}_{ci} = k(i)x_{ci} + c_a(i)\dot{x}_{ci} \tag{39}$$

or, in vector form:

$$\mathbf{t}_c = \operatorname{diag}(\mathbf{x}_c)\mathbf{k} + \operatorname{diag}(\dot{\mathbf{x}}_c)\mathbf{c}_a \tag{40}$$

where  $\mathbf{k} = [k(1), k(2), k(3), k(4)]^T$  is the constant vector of cable stiffnesses and  $\mathbf{c}_a = [c_a(1), c_a(2), c_a(3), c_a(4)]^T$  is the constant vector of damping coefficients. Using Eq. (29) and the friction model, the vector of cable tensions can be written as

$$\mathbf{t}_{c} = \mathbf{f} - \operatorname{diag}\left(\frac{2}{\pi}\arctan\left(c_{s}\dot{\theta}\right)\right)\mathbf{f}_{s} - \frac{R}{r_{d}}I_{a}\ddot{\theta}$$
 (41)

Combining Eqs. (40) and (41), the following vector equation is obtained:

$$\operatorname{diag}(\mathbf{x}_c)\mathbf{k} + \operatorname{diag}(\dot{\mathbf{x}}_c)\mathbf{c}_a = \mathbf{f} - \operatorname{diag}\left(\frac{2}{\pi}\arctan\left(c_s\dot{\theta}\right)\right)\mathbf{f}_s - \frac{R}{r_d}I_a\ddot{\theta}$$
 (42)

This vector equation provides four equations that are linear in the friction, stiffness, and damping coefficients of each cable.

5.1.3 Complete Linear System. Equations (38) and (42), which describe the contribution of friction in the dynamic model and in the cable stiffness model, respectively, are now gathered to obtain a system of equations linear in the parameters to identify:

$$\mathbf{D}\mathbf{v} = \mathbf{y} \tag{43}$$

where **v** if the vector of  $n_p = 12$  parameters to identify:  $\mathbf{v} = \begin{bmatrix} \mathbf{k} & \mathbf{c}_a & \mathbf{f}_s \end{bmatrix}^T$ . Now, the contribution of friction in Eqs. (38) and (42) is not of the same order because the coefficients of **Z** are of the order of  $10^{-2}$ . To avoid any discrepancies, Eq. (38) is multiplied by a weight W = 100.

Vector  $\mathbf{y}$  is given by

$$\mathbf{y} = \begin{pmatrix} \mathbf{f} - \frac{R}{r_d} I_a \ddot{\theta} \\ W(\mathbf{Z}(\mathbf{f} + I_a \left(\frac{R}{r_d}\right)^2 \ddot{\mathbf{x}}_c) - \mathbf{M}\ddot{\alpha} - \mathbf{g} - \mathbf{c}) \end{pmatrix}$$
(44)

and D is written as

$$\mathbf{D} = \begin{pmatrix} \operatorname{diag}(\mathbf{x}_c) & \operatorname{diag}(\dot{\mathbf{x}}_c) & \operatorname{diag}\left(\frac{2}{\pi}\arctan\left(c_s\dot{\theta}\right)\right) \\ \mathbf{0}_{3\times4} & \mathbf{0}_{3\times4} & W\mathbf{Z}\operatorname{diag}\left(\frac{2}{\pi}\arctan\left(c_s\dot{\theta}\right)\right) \end{pmatrix}$$
(45)

Friction parameters are also identified independently for the purpose of comparison with the global identification. In this case, there are only  $n_p = 4$  parameters to identify,  $\mathbf{v} = \mathbf{f}_s = [f_s(1), f_s(2), f_s(3), f_s(4)]^T$  and

$$\mathbf{y} = \mathbf{f} - \frac{R}{r_d} I_a \ddot{\theta},\tag{46}$$

$$\mathbf{D} = \operatorname{diag}\left(\frac{2}{\pi}\arctan\left(c_s\dot{\theta}\right)\right) \tag{47}$$

**5.2 Identification Method.** The motor forces  $\mathbf{f}$  and motor angular positions  $\theta$  as well as the joints orientation angles  $\alpha$  are sampled along a specific trajectory to obtain r sampled values. Then,  $\mathbf{v}$  can be estimated as the least square solution  $\hat{\mathbf{v}}$  of the linear system:

$$\mathbf{W}\mathbf{v} + \rho = \mathbf{Y} \tag{48}$$

where **W** is a  $(r \times n_p)$  observation matrix defined by a sampling of the regressor **D** defined in Eq. (45) or (47), **Y** is a sampling of **y** defined in Eqs. (44) or (46),  $\rho$  is the vector of residues or vector of errors. The number r of samples must be greater than the number of parameters to identify:  $r > n_p$  [39].

The parameter vector  $\mathbf{v}$  minimizing the 2-norm of  $\rho$  is given by

$$\mathbf{v} = \mathbf{W}^{+} \mathbf{Y} \tag{49}$$

The quality of the solution depends on the rank of the condition number of W which depends on the choice of the trajectory during the identification process. If the measurement noises are independent in W and Y, then the quality of the identification can be evaluated with the covariance matrix  $C_X$ :

$$\mathbf{C}_X = \sigma_o^2 (\mathbf{W}^T \mathbf{W})^{-1} \tag{50}$$

where  $\sigma_{\rho}^2 = \frac{\|\mathbf{Y} - \mathbf{W}\mathbf{v}\|^2}{r - N_p}$ . The standard deviation of the *i*th parameter is  $\sigma_i = \sqrt{\mathbf{C}_X(i,i)}$  and the confidence interval is given by  $2\sigma_i$ . The relative standard deviation  $100\frac{\sigma_i}{\mathbf{v}_i}$  evaluates the quality of identification of parameter *i*.

5.2.1 Filtering Aspects. The data used for the identification are the measurements of the motor angular positions  $\theta$ , the input actuation forces  $\mathbf{f}$  (assuming that the electrical response of the motors is perfect), and the manipulator configuration  $\alpha$  evaluated from the image processing. Cable elongations are deduced from Eq. (27), after an initialization in straight position. The derivatives of the measured quantities are calculated by centered numerical differentiation after filtering.

It is essential to filter the measured data to avoid bias arising from noisy discrete measurements [41]. A lowpass Butterworth filter is used to filter the measurements in forward and reverse direction with the MATLAB filtfilt function. The derivatives of  $\theta$  are obtained with a central difference algorithm to avoid phase shift. The cutoff frequency of the Butterworth filter is computed with the method explained in Ref. [37].

The video frequency used to measure the orientation of the X-joints orientation is 30 Hz while the frequency of the control law is 500 Hz. To synchronize the data, the motor encoder measurements are resampled with the "decimate" MATLAB procedure to 31.25 Hz, the value closest to 30 Hz. The data from video are also resampled to 31.25 Hz with a linear interpolation.

5.2.2 Identification Trajectory. An excitation trajectory must be designed for the identification in order to improve estimation accuracy [42]. Regarding cable elasticity, the trajectory must be executed with a large exploration of applied forces since cable tensions depend mostly on the latter. Regarding friction, the trajectory must be chosen in order to meet the largest set of possible values of  $\mathbf{Z}$  ( $\alpha$ ). Accordingly, the identification trajectory is defined so that it meets a large set of configurations  $\alpha$  and thus requests high forces. Symmetric motions with respect to the straight configuration are considered, knowing that the actuation differs on the right and left sides. The identification trajectory is composed of four successive sequences. Each sequence is composed of the following six elementary motions: starting from its straight configuration,

Table 1 Identified parameters

	Ident. with Eqs. (44), (45)	Ident. with Eqs. (46), (47)	Stiffness		Damping	
$f_s(1)(N)$	14.42	14.13	k(1) (N m <sup>−1</sup> )	10,454	$c_a(1) \text{ (N s m}^{-1})$	805
Relative $\sigma_i$	1.68%	2.52%	Relative $\sigma_i$	1.28%	Relative $\sigma_i$	30.5%
$f_s(2)(N)$	17.65	17.70	$k(2) (N m^{-1})$	9256	$c_a(2)  (\text{N s m}^{-1})$	333
Relative $\sigma_i$	1.38%	2.02%	Relative $\sigma_i$	1.37%	Relative $\sigma_i$	68.7%
$f_s(3)(N)$	18.77	18.97	$k(3) (N m^{-1})$	8319	$c_a(3)  (\text{N s m}^{-1})$	-218
Relative $\sigma_i$	1.32%	1.88%	Relative $\sigma_i$	1.60%	Relative $\sigma_i$	94.3%
$f_s(4)(N)$	16.18	16.29	$k(4) (\text{N m}^{-1})$	5336	$c_a(4)  (\text{N s m}^{-1})$	370
Relative $\sigma_i$	0.88%	1.25%	Relative $\sigma_i$	1.85%	Relative $\sigma_i$	64.3%

the manipulator moves to a desired  $\alpha^d$ , moves backward to the opposite configuration  $-\alpha^d$ , goes back to the straight configuration, moves to  $-\alpha^d$ , moves backward to  $\alpha^d$  and finally goes back to the straight configuration. For the first sequence, all joints move in the same direction and  $\alpha^d = [40 \, \mathrm{deg}, 30 \, \mathrm{deg}, 20 \, \mathrm{deg}]^T$ . For the second (resp. third, fourth) sequence, the first (resp. second, third) joint moves in the opposite direction to that of the other joints:  $\alpha^d = [-40 \, \mathrm{deg}, 30 \, \mathrm{deg}, 20 \, \mathrm{deg}]^T$ ,  $[40 \, \mathrm{deg}, -30 \, \mathrm{deg}, 20 \, \mathrm{deg}]^T$ ,  $[40 \, \mathrm{deg}, 30 \, \mathrm{deg}, -20 \, \mathrm{deg}]^T$ , respectively. All desired joint configurations are connected with cycloidal functions. A closed loop control is used to follow the identification trajectory. The identification process being not affected by the performance of the control, the closed loop is simply tuned to produce smooth actuation force variations.

Video recording of the identification trajectory can be found online.<sup>2</sup>

5.3 Results With Linear Cable Stiffness. The identification of friction is conducted with two different models: the complete model defined by Eqs. (44) and (45), and the model without cable elasticity defined by Eqs. (46) and (47). Results are given in Table 1. According to the relative standard deviation, the friction and stiffness coefficients are well identified. The friction parameters are similar whether identified with the elastic cable model or not. The first three friction coefficients are associated with the three cables pulling on the right. It is apparent that  $f_s(3) > f_s(2) > f_s(1)$ , which is due to pulley friction. Indeed, the number of pulleys met by cables increases with the cable index and so does friction. Since the contribution of  $diag(\dot{\mathbf{x}}_c)c_a$  in Eq. (42) is limited, the damping coefficients are not well identified as shown by the standard deviation. In fact, a variation of the damping coefficients  $c_{ai}$ from 0 to 3000 has a limited impact on the other coefficients, with a variation less than 5%. In the following, a large value is chosen for all damping coefficients in order to damp the elongation in our model:  $c_{ai} = 3000$ , i = 1, ..., 4.

We want to assess the quality and interest of identification. The identification model relies on Eqs. (38) and (42). Thus, we verify if these equations are well satisfied with the identified model. We also want to show how the identified model improves results with respect to a model with no friction and no cable elongation.

Equations (38) can be rewritten as follows:

$$\mathbf{Z}(\alpha)\mathbf{f} = \mathbf{Z}(\alpha) \left( \operatorname{diag} \left( \frac{2}{\pi} \arctan \left( c_s \dot{\theta} \right) \right) \mathbf{f}_s - I_a \left( \frac{R}{r_d} \right)^2 \ddot{\mathbf{x}}_c \right) + \mathbf{M}(\alpha) \ddot{\alpha} + \mathbf{c}(\alpha, \dot{\alpha}) + \mathbf{g}(\alpha)$$
(51)

Let  $\Gamma_m$  be the torque vector calculated from motor forces:  $\Gamma_m = \mathbf{Z}(\alpha)\mathbf{f}$ . Let  $\Gamma_{rf}$  (resp.  $\Gamma_r$ ) be the torque vector reconstructed with the right-hand side (RHS) of Eq. (51) with friction (resp. with no

friction):

$$\Gamma_{rf} = \mathbf{Z}(\alpha) \left( \operatorname{diag} \left( \frac{2}{\pi} \arctan \left( c_s \dot{\theta} \right) \right) \mathbf{f}_s - I_a \left( \frac{R}{r_d} \right)^2 \ddot{\mathbf{x}}_c \right) + \mathbf{M}(\alpha) \ddot{\alpha} + \mathbf{c}(\alpha, \dot{\alpha}) + \mathbf{g}(\alpha)$$
(52)

$$\Gamma_r = \mathbf{Z}(\alpha) \left( -I_a \left( \frac{R}{r_d} \right)^2 \ddot{\mathbf{x}}_c \right) + \mathbf{M}(\alpha) \ddot{\alpha} + \mathbf{c}(\alpha, \, \dot{\alpha}) + \mathbf{g}(\alpha)$$
 (53)

Figure 4 shows the plots of  $\Gamma_m$ ,  $\Gamma_{rf}$ , and  $\Gamma_r$ . The goal is to highlight the contribution of the friction terms and to show that these terms must be taken into account. The values of the friction parameters identified with or without the elastic model (e.g., in the first and second column of Table 1) being very close, the corresponding plots in Fig. 4 cannot be distinguished. It is apparent that the friction model improves reconstruction of  $\mathbf{Z}(\alpha)\mathbf{f}$ . The mean error between the measured values and the values evaluated by the model is  $0.18 \,\mathrm{Nm\,rad^{-1}}$  when friction is considered but it is  $0.55 \,\mathrm{Nm\,rad^{-1}}$ without friction. Friction also changes the shape of the curves, which the model does effectively. Some errors remain, in particular at low velocities, for which our model is not efficient. More precise friction models [36] or improved control schemes [7] could be used in future. It can be noticed that antagonistically actuated systems are very sensitive to friction. When changing the orientation of a X-joint, one cable winds up when the other unwinds, and the torques produced by friction via matrix **Z** add up.

Cable tension can be computed in two ways. The first one is from the forces using the RHS of Eq. (42). The second one is from the stiffness and cable elongation coefficients using the LHS of Eq. (42). Figure 5 shows the corresponding plots. The same shape is observed, but significant differences exist for high tension values. For a better visualization of this phenomenon, the reconstruction error obtained as the absolute difference between the cable tensions calculated with the RHS and LHS of Eq. (42) is plotted against the applied forces in Fig. 6. The reconstruction error clearly increases with cable tension, which suggests that cable stiffness increases with tension. Accordingly, we propose in section 5.4 a new model for the cable elongation, with a stiffness coefficient that varies linearly with tension.

**5.4 Cable Stiffness Depending on Tension.** Non-linear cable models, as in Ref. [43], were also tested, but the identification did not significantly improve the results compared to the linear models. Therefore, we developed an ad hoc model to fit the observed measurements. The following new cable stiffness model is proposed:

$$t_{ci} = (k(i) + \lambda(i)t_{ci})x_{ci} + c_a(i)\dot{x}_{ci}$$
 (54)

where  $\lambda = [\lambda(1), \lambda(2), \lambda(3), \lambda(4)]^T$  is an additional vector to be identified.

<sup>&</sup>lt;sup>2</sup>https://uncloud.univ-nantes.fr/index.php/s/8wzgsPi65cxDxG9

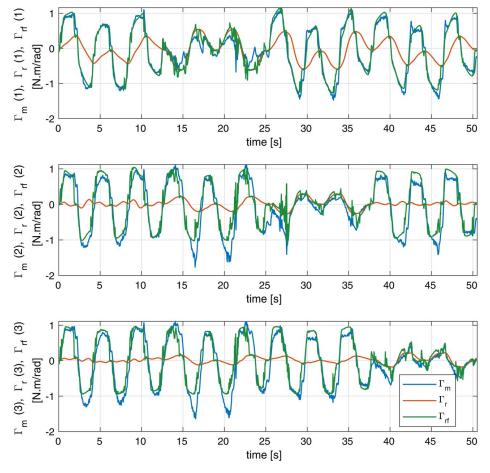


Fig. 4 Time history of torques calculated from the dynamic model with friction ( $\Gamma_{rf}$ , green plot), without friction ( $\Gamma_r$ , red plot) and calculated from motor forces ( $\Gamma_m$ , blue plot) (Color version online.)

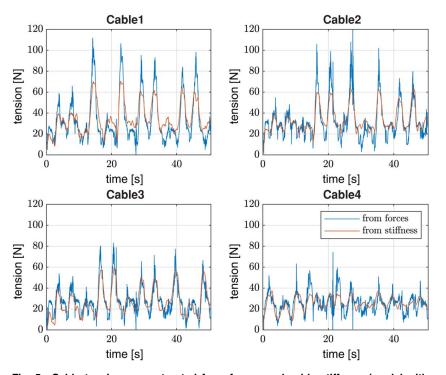


Fig. 5 Cable tension reconstructed from forces and cable stiffness (model with constant stiffness)

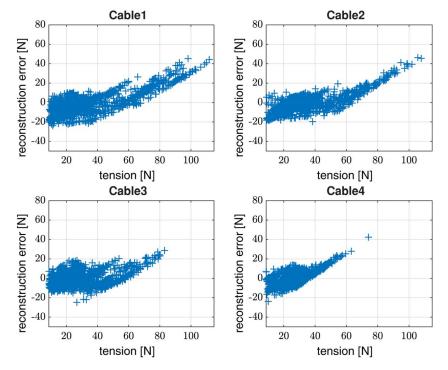


Fig. 6 Reconstruction error as a function of the tension computed from the forces (model with constant stiffness)

Tension in cable i can also be calculated by the  $i^{th}$  line of Eq. (41):

$$t_{ci} = f_i - \frac{2}{\pi} \arctan(c_s \dot{\theta}_i) f_s(i) - \frac{R}{r_d} I_a \ddot{\theta}_i$$
 (55)

If Eq. (54) is reported into Eq. (55), we obtain a model that is nonlinear in the parameters  $\lambda(i)$  and  $f_s(i)$  to identify. In order to build a linear model and since the friction coefficients can be identified separately, the friction parameter values  $\mathbf{f_s}$  identified previously can be used. In this case,  $t_{ci}$  can be calculated using Eq. (55) and the measurements. The identification model is built using Eq. (54). It can then be written as Eq. (43) where  $\mathbf{v}$  if the vector of  $n_p = 12$  parameters to identify:  $\mathbf{v} = [\mathbf{k} \ \lambda \ \mathbf{c}_a]^T$ . Vector  $\mathbf{y}$  is defined by

$$y = t_c \tag{56}$$

and D is

$$\mathbf{D} = \left( \operatorname{diag}(\mathbf{x}_c) \quad \operatorname{diag}(\mathbf{xt}) \quad \operatorname{diag}(\dot{\mathbf{x}}_c) \right) \tag{57}$$

where  $\mathbf{xt} = [x_{c1} \ t_{c1}, x_{c2} \ t_{c2}, x_{c3} \ t_{c3}, x_{c4} \ t_{c4}]^T$ .

The same method as previously described is used for identification, and the identified parameters are given in Table 2. The stiffness parameters are well identified but the damping coefficients

Table 2 Identified parameters with the new cable stiffness model (variable stiffness)

Stiffness		Stiffness		Damping	
$k(1) (N m^{-1})$	5378	$\lambda(1) \ (m^{-1})$	103.3	$c_a(1) \text{ (N s m}^{-1})$	516.8
Relative $\sigma$	1.6%	Relative $\sigma$	1.5%	Relative $\sigma$	14.5%
$k(2) (\text{N m}^{-1})$	5364	$\lambda(2) \ (\text{m}^{-1})$	88.7	$c_a(2) \text{ (N s m}^{-1})$	185.9
Relative $\sigma$	1.7%	Relative $\sigma$	2.1%	Relative $\sigma$	37.8%
$k(3) (\text{N m}^{-1})$	6231	$\lambda(3) \ (\text{m}^{-1})$	53.2	$c_a(3) \text{ (N s m}^{-1})$	-273.3
Relative $\sigma$	1.7%	Relative $\sigma$	4.8%	Relative $\sigma$	22.9%
$k(4) (\text{N m}^{-1})$	1908	$\lambda(4) \ (\text{m}^{-1})$	120.9	$c_a(4) \text{ (N s m}^{-1})$	241.5
Relative $\sigma$	5.0%	Relative $\sigma$	2.6%	Relative $\sigma$	30.7%

are, again, badly identified. The components of the stiffness vector  $\mathbf{k}$  are lower than the values obtained in Table 1 but they increase with tension since parameters  $\lambda$  are positive.

Figure 7 shows the reconstruction of cable tension, calculated in two ways like in Fig. 5. The results are improved as compared to Fig. 5 and tension is well modeled even for high values. The mean error is now 4.75 N, against 6.88 N with the previous model. The reconstruction error does not increase with the forces.

5.5 Reconstruction of  $\hat{\alpha}$  Based on Cable Stiffness Model. Now that stiffness is identified,  $\hat{\alpha}$  can be evaluated online for control. It can be reconstructed using the elastic model and the measurements of  $\theta$ . Reconstruction precision is evaluated by comparison with the measurement of  $\alpha$  by image processing :  $\alpha_{IP}$ . Cable elongation  $\mathbf{x}_c$  is estimated with the elastic model. For a short time-step  $\Delta t$ , a discrete prediction of the cable elongation by integration of Eq. (54) is

$$\mathbf{x}_{c}(t + \Delta t) = \mathbf{x}_{c}(t) + \frac{\Delta t}{\mathbf{c}_{a}} \left( \mathbf{f}(t) - \mathbf{f}_{s} \frac{2}{\pi} \arctan\left(c_{s} \dot{\theta}(t)\right) - (\mathbf{k} + \lambda \mathbf{f}(t)) \mathbf{x}_{c}(t) \right)$$
(58)

Considering an initial zero elongation,  $\mathbf{x}_c$  can be updated at each time-step (a sampling period of 2 ms is used). In the following, three cable models are considered: (i) inelastic cable:  $\mathbf{x}_c = 0$ , (ii) model 1: cable with constant stiffness ( $\lambda = 0$  and  $\mathbf{k}$  given in Table 1), and (iii) model 2: cable with variable stiffness ( $\lambda$  and  $\mathbf{k}$  given in Table 2). Then,  $\hat{\alpha}$  can be computed with only three cable data using Eq. (33) and can also be computed with the four cable data using Eq. (35).

Figure 8 shows the reconstruction  $\hat{\alpha}$  and its values obtained via image processing  $\alpha_{IP}$ . When  $\hat{\alpha}$  is computed from Eq. (33), the mean error without model is 7.87 deg against 1.67 deg with the constant stiffness model and 1.51 deg with the variable stiffness model. When  $\hat{\alpha}$  is computed from Eq. (35), the mean error without model is 2.79 deg against 1.84 deg with the constant stiffness model, and 1.36 deg with the variable stiffness model. The model of elastic cable gives a better evaluation  $\hat{\alpha}$  than the model

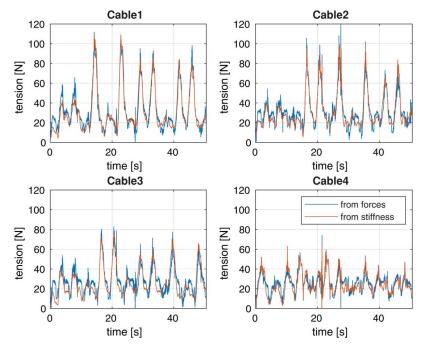


Fig. 7 Cable tension reconstructed from forces and cable stiffness (variable stiffness)

with no elasticity. The use of four cable data with Eq. (33) is preferable and will be chosen for control purposes.

**5.6 Simulator.** One of the main objectives of the identification is to have a simulator of the manipulator with a behavior close to the prototype, in order to prepare the experiments in good conditions. Our manipulator is force-controlled, i.e., the inputs are the forces transmitted by the motors to the cables. The role of the simulator

is to predict the behavior of the manipulator and, more importantly, the evolution of  $\alpha$  and  $\theta$ , and thus of  $x_c$  during motion. The state variables for the simulator are  $\alpha$ ,  $\theta$ ,  $\dot{\alpha}$ , and  $\dot{\theta}$ . The model is based on Eqs. (32) and (29). Cable elongation is calculated by Eq. (27) and its derivatives. Cable tension is calculated using Eq. (54).

A comparison study is conducted between (i) results obtained with the simulator considering no friction and no cable elasticity, (ii) results obtained with the simulator derived from the identification work, and (iii) experimental results. The comparison is done

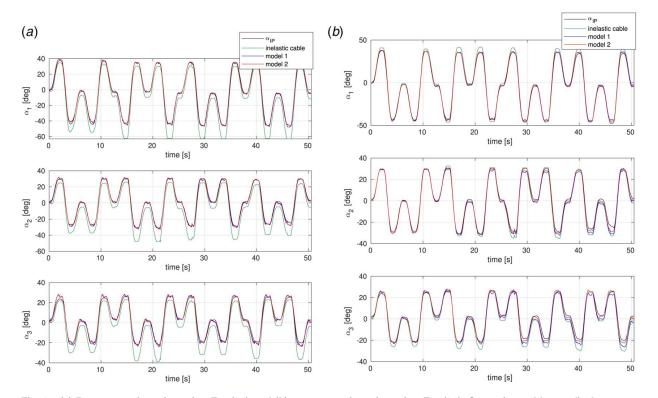


Fig. 8 (a) Reconstruction of  $\alpha$  using Eq. (33) and (b) reconstruction of  $\alpha$  using Eq. (35). Green (resp. blue, red) plot corresponds to inelastic cable model (resp. cable model 1, cable model 2). Dark plot shows  $\alpha$  measured by image processing.

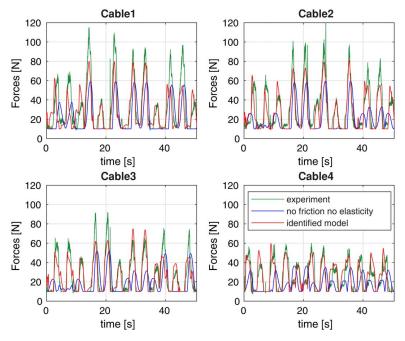


Fig. 9 Comparison of the applied forces in experiment and in simulation

with the trajectory and control law used for the identification. The control law will be detailed in Sec. 6 as test1.

Forces applied during experiment are compared with the forces obtained in simulation in Fig. 9. Friction and elastic models clearly improve the efficiency of the simulator. Force peaks are better approached. Some of them were not present with the simulator without friction model. Thus, the new simulator is more efficient to prepare our experiments. However, some peaks remain lower with the simulator based on identified parameters, than in experiment. This is due to the identification of friction that is not perfect. Regarding  $\alpha$  and  $\theta$ , plots of their time histories are very close for simulations and experiments. Quantification of the differences is given in Table 3.

In the calculation of cable lengths, we have paid particular attention to the effect of the pulleys, whereas their effect was neglected in Ref. [9]. We will now quantify their effects. First of all, we note that pulleys introduce a coupling between the joints: forces exerted by the short cables induce a motion on the lower X-joints. This was not the case when pulleys were neglected, a situation where  $\mathbf{Z}_1$  was diagonal.

Three indicators are used here to evaluate the pulley contribution. The first one is the torques applied on X-joints: we calculate the norm of the difference between the torques calculated with and without pulleys, for a large number of random joint configurations  $-\frac{\pi}{2} < \alpha_i < \frac{\pi}{2} \text{ and applied forces } f_{\min} < f_i < f_{\max}.$  In average, we obtain a difference of 16%.

Table 3 Mean value for the errors of the estimation  $\mathbf{e}_{\alpha}=\alpha-\hat{\alpha}$  and tracking error e in simulations and in experiment

Control law	Simulation with rigid cable model	Simulation with identified cable model	Experiment
With pulley	$e_{\alpha} = 0.35 \deg$ $e = 0.08 \deg$	$e_{\alpha} = 1.68 \deg$ $e = 0.11 \deg$	$e_{\alpha} = 2.69 \deg$ $e = 0.64 \deg$
Without pulley	$e_{\alpha} = 2.00 \deg$ $e = 0.21 \deg$	$e_{\alpha} = 2.88 \deg$ $e = 0.26 \deg$	

Note: The experiments are only run with a control including pulleys.

As a second indicator, the impact on the computation of the orientation of the X-joints  $\alpha$  as a function of the position of the encoders  $\theta$  using Eq. (35) is evaluated. A statistical study for different configurations of the robot  $\alpha(t)$  and different displacements of the motor  $\theta(t + \Delta t) - \theta(t)$  shows a variation of 13% on average on  $\alpha(t + \Delta t) - \alpha(t)$ .

The last indicator is evaluated in simulation. The trajectory used for the identification is simulated with two different controls (detailed in Sec. 6. The two controls consider the friction model and use the pseudoinverse computation of  $\alpha$  from  $\theta$  neglecting cable elasticity. In the first control, we consider the real radius of the pulleys, while in the second control, the radius of the pulleys is neglected (r=0 mm). In the two controls, thus, the computation of **Z** is not the same, which changes the computation of both the configuration  $\alpha$  and the forces f. In robot model used for the simulation, pulleys as well as friction and cable elasticity are taken into account. The first control produces a mean error on the estimation of  $\alpha$  of 1.65 deg, versus 2.88 deg for the second control. Moreover, the mean tracking error (difference between measured  $\alpha$  and desired  $\alpha^d$ ) is 0.12 deg on  $\alpha$  for the first control, versus 0.32 deg for the second control. Results are summarized in Table 3. It is observed that not taking into account pulleys in the control model creates an error that is still greater than not taking into account cable flexibilities and friction in the simulations.

In summary, pulleys have a significant impact on control and must not be neglected.

# 6 Control

**6.1** Computed Torque Control Law. Three control laws are proposed. The first one was used to run the identification trajectory on the prototype. It is now used for the comparison analysis. The objective of the control law is to define the force  $\mathbf{f}$  applied by the motor to track a desired trajectory  $\alpha^d(t)$  in joint space. A PID controller is used to define the desired joint acceleration  $\mathbf{w}(t)$  in closed loop as follows:

$$\mathbf{w}(t) = \ddot{\alpha}^d(t) + k_v \dot{\mathbf{e}}(t) + k_p \mathbf{e}(t) + k_i \int_{t'=0}^t (\mathbf{e}(t')) dt$$
 (59)

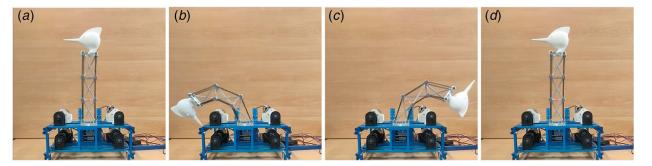


Fig. 10 The main configurations of the trajectory for the control test at (a) 0 s, (b) 4 s, (c) 7 s, and (d) 10 s

where  $\mathbf{e} = \alpha^d - \hat{\alpha}$  The gains are:  $k_i = \omega^3$ ,  $k_v = 3\omega$ , and  $k_p = 3\omega^2$  and we set  $\omega = 11$  for all experiments. Equation (59) can be achieved if the controlled forces satisfy the dynamic model in Eq. (32), namely if

> $\mathbf{M}(\alpha)\mathbf{w}(t) + \mathbf{c}(\alpha, \dot{\alpha}) + \mathbf{g}(\alpha)$  $= \mathbf{Z}(\alpha) \left( \mathbf{f} - \mathbf{f}_{friction}(\dot{\theta}) + I_a \left( \frac{R}{r_d} \right)^2 \ddot{\mathbf{x}}_c \right)$ (60)

Now the following points have to be considered:

- Since the manipulator is over-actuated, infinitely many forces can produce a desired acceleration, but in the same time, cable tensions must remain positive.
- There is no sensor to measure  $x_c$  and  $\alpha$ , they have to be reconstructed online from the identification models developed
- The online computation of the dynamic model is a bit cumbersome and would become intractable if several X-joints are added. Accordingly, control laws will be designed with a simplified dynamic model.

6.2 Over-Actuation and Choice of Cable Tensions. Vector f appears in Eq. (60) via  $\mathbf{Z}(\alpha)\mathbf{f}$  where  $\mathbf{Z}$  is a  $3\times4$  matrix and  $\mathbf{f}$  is a

vector of dimension 4. Let us define

$$\Gamma^{d} = \mathbf{M}(\alpha)\mathbf{w}(t) + \mathbf{c}(\alpha, \dot{\alpha}) + \mathbf{g}(\alpha) + \mathbf{Z}(\alpha) \left(\mathbf{f}_{friction}(\dot{\theta}) - I_{a} \left(\frac{R}{r_{d}}\right)^{2} \ddot{\mathbf{x}}_{c}\right)$$
(61)

Then, Eq. (60) is rewritten as  $\Gamma^d = \mathbf{Z}(\hat{\alpha})\mathbf{f}$  and, using Eq. (23), as

$$\Gamma^d = \mathbf{Z}_1(\hat{\alpha})\mathbf{f}_{13} + \mathbf{Z}_2(\hat{\alpha})f_4 \tag{62}$$

where  $\mathbf{f}_{13} = [f_1, f_2, f_3]^T$ . Since  $\mathbf{Z}_1$  is triangular and invertible,  $\mathbf{f}$  can be expressed as a function of  $f_4$  (the force applied on the long cable) as follows:

$$\mathbf{f} = \begin{pmatrix} \mathbf{Z}_1(\hat{\alpha})^{-1} \Gamma^d \\ 0 \end{pmatrix} + \begin{pmatrix} -\mathbf{Z}_1(\hat{\alpha})^{-1} \mathbf{Z}_2(\hat{\alpha}) \\ 1 \end{pmatrix} f_4$$
 (63)

There are thus infinitely many forces  $\mathbf{f}$  parametrized by  $f_4$ . Two constraints are proposed to select an appropriate solution for control.

The first constraint is to impose a minimal tension  $t_{\min} > 0$  to avoid slack cables. From Eq. (29), this condition yields:

$$f_i \ge t_{\min} + \frac{2}{\pi} \arctan(c_s \dot{\theta}_i) f_s(i) + \frac{R}{r_d} I_a \ddot{\theta}_i$$
 (64)

Since our friction model is not accurate at rest, an additional condition is considered for more safety:  $f_i \ge f_{\min} > 0$ . Altogether, each

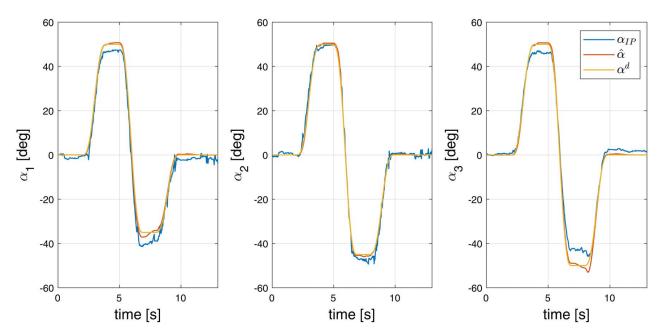


Fig. 11 Time history of joint angles during test 1 (friction and cable elasticity neglected)

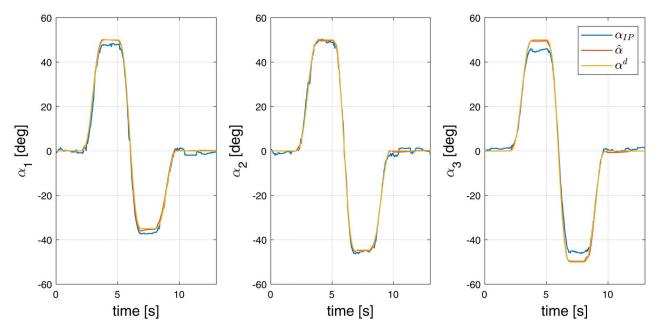


Fig. 12 Time history of joint angles during test 2 (friction is considered but not cable elasticity)

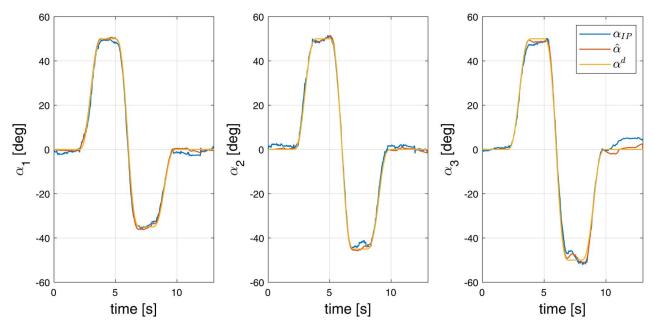


Fig. 13 Time history of joint angles during test 3 (both friction and cable elasticity are considered)

cable force  $f_i$ ,  $1 \le i \le 4$ , must satisfy:

$$f_i \ge max \left( t_{min} + \frac{2}{\pi} \arctan\left( c_s \dot{\theta}_i \right) f_s(i) + \frac{R}{r_d} I_a \ddot{\theta}_i, f_{min} \right)$$
 (65)

In all experiments, we set  $f_{min} = 2 \text{ N}$  and  $t_{min} = 8.5 \text{ N}$ . These values have been chosen experimentally as the minimal values for which no cable is slack.

The second constraint is to minimize the current consumption of the actuators. Accordingly, the norm of  $\mathbf{f}$  is minimized. Due to the antagonist actuation, this constraint induces a minimal force in at least one cable. The applied force is calculated with the following method. Four solutions  $f_4$  to Eq. (63) are computed such that  $f_i = \begin{pmatrix} 2 & \cdots & R & \cdots & R \end{pmatrix}$ 

$$\max\left(t_{min} + \frac{2}{\pi}\arctan\left(c_{s}\dot{\theta}_{i}\right)f_{s}(i) + \frac{R}{r_{d}}I_{a}\ddot{\theta}_{i}, f_{min}\right) \text{ and the greatest}$$

solution is chosen. This choice ensures that all the conditions are satisfied since the components of  $-\mathbf{Z}_1^{-1}\mathbf{Z}_2$  are all positive, whatever the value of  $\alpha$ . If the resulting force  $f_i$  is greater than  $f_{max}$ , we take  $f_i = f_{max}$ . In practice, because the measured values  $\theta_i$  and  $\dot{\theta}_i$  are noisy, they are replaced by the desired values  $\theta_i^d$  and  $\dot{\theta}_i^d$  in Eq. (65) to avoid any force discontinuity.

The control law developed for the manipulator at hand can be extended to manipulators with any number N of modules equipped with one long cable on one side and N additional cables on the other side

**6.3** Three Control Laws. A computed torque control law can be built with  $\Gamma^d$  defined by Eq. (61). To reduce the calculation

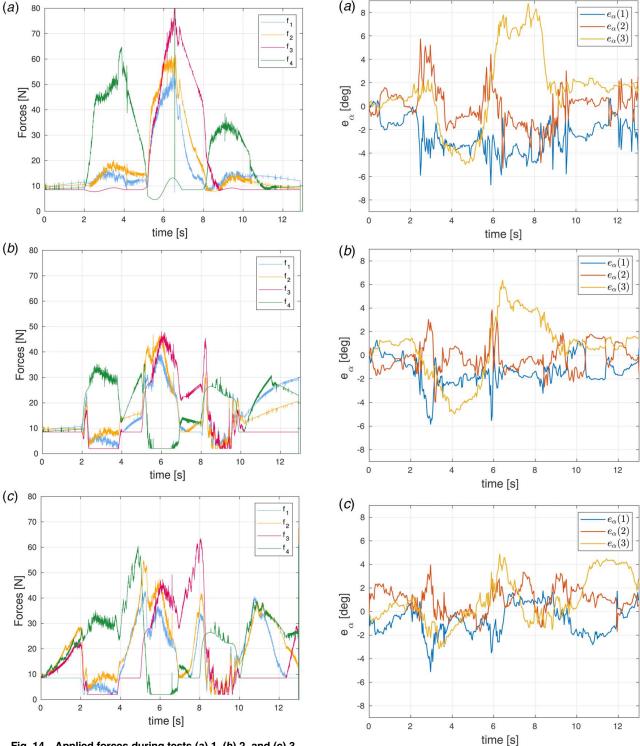


Fig. 14 Applied forces during tests (a) 1, (b) 2, and (c) 3

burden, however, a simplified dynamic model is used. Since the manipulator is light and experienced motions are not very fast, the centrifugal terms as well as the acceleration of the cable elongation can be neglected. Finally, the measurement  $\theta$  is replaced by the desired value  $\theta^d$  to avoid variations in the friction term at low velocities:

$$\Gamma^{d} = \mathbf{M}(\hat{\alpha})\mathbf{w}(t) + \mathbf{g}(\hat{\alpha}) + \mathbf{Z}(\hat{\alpha})(\mathbf{f}_{friction}(\dot{\boldsymbol{\theta}}^{d}))$$
 (66)

The orientation angles of the X-joints  $\hat{\alpha}$  is calculated with Eq. (35) from the measurements  $\theta$ . Three different tests are

Fig. 15 Error  $\mathbf{e}_{\alpha}$  between  $\hat{\alpha}$  (computed with pseudo inverse) and  $\alpha_{IP}$  (obtained from video) during tests (a) 1, (b) 2, and (c) 3

carried out to evaluate the efficiency of the new control, corresponding to the following three control laws:

- test 1: friction and cable elasticity are neglected;
- test 2: friction is taken into account but cable elasticity is neglected;
- test 3: both friction and cable elasticity are taken into account.

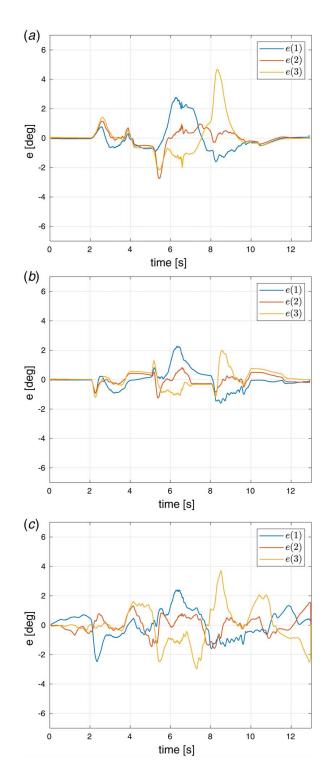


Fig. 16 Error e between  $\hat{\alpha}$  (computed with pseudo inverse) and the desired trajectory  $\alpha^d$  during tests (a) 1, (b) 2, and (c) 3

# 7 Control Tests Analysis

The trajectory used for the control tests involves a motion in the same direction for the three X-joints. This movement seems to be the most interesting to access a large workspace with reasonable forces. Starting from straight configuration, the manipulator moves to  $\alpha^d = [50 \deg, 50 \deg, 50 \deg]^T$ , goes back to the straight configuration, then moves to  $\alpha^d = [-30 \deg, -40 \deg, -50 \deg]^T$  and finally goes back to straight configuration. All desired joint configurations are connected with cycloidal functions. The four main configurations of the trajectory are shown in Fig. 10.

Figure 11 shows the results of test 1 (model with no friction and inelastic cables). Time histories of X-joint orientation angles are plotted for three different values:  $\alpha^d$ , which is the desired trajectory,  $\hat{\alpha}$ , the estimated orientation and  $\alpha_{IP}$ , the orientation computed offline from video and image processing. The results are fairly good thanks to the computation of  $\hat{\alpha}$  with the pseudo-inverse: the mean absolute difference between  $\hat{\alpha}$  and  $\alpha^d$  is about 0.52 deg. The mean absolute difference between  $\hat{\alpha}$  and  $\alpha_{IP}$  is about 2.19 deg, which shows that joint angles are not perfectly evaluated. While the tracking error are small, the extreme positions are not well reached since the reconstruction of  $\hat{\alpha}$  is not precise enough, especially for joints 1 and 3.

Figure 12 shows the results of test 2 (model with friction and inelastic cables). As expected, results are better than those of the first test. The mean absolute difference between  $\hat{\alpha}$  and  $\alpha^d$  (resp. between  $\hat{\alpha}$  and  $\alpha_{IP}$ ) is about 0.37 deg (resp. 1.51 deg). The better prediction of the feed-forward term by the knowledge of the dry friction terms, allows reducing the contribution of the feedback term and reduces tracking errors. In addition, the friction model allows applying lower forces. Less energy is thus required and cable elongation is reduced, which leads to a better estimation  $\hat{\alpha}$ . The extreme positions are better reached than with the previous model since the estimation  $\hat{\alpha}$  is slightly improved, but errors are visible on joints 1 and 3.

Figure 13 shows the results of the last test (model with both friction and cable elasticity). The improvement of the estimation  $\hat{\alpha}$  produces a reduction of the mean absolute difference between  $\hat{\alpha}$  and  $\alpha_{IP}$ , which is now about 1.09 deg. The extreme positions are also better reached than in the other experiments. However, the mean absolute difference between  $\hat{\alpha}$  and  $\alpha^d$  is about 0.66 deg, which is a bit greater than in the previous experiments. This increase in the tracking errors might be due to the computed control law that resorts to Eq. (66) where cable elasticity is neglected. A control law based on a dynamic model with elastic cables could have been defined. However, such a model would require calculation of the derivatives of the cable elongation, which would add noise.

To finish, we now compare the forces and errors for the three control tests. Figure 14 shows the applied forces during the three experiments. The reference trajectory includes displacement phases and stop phases. In stop phases, the condition on minimum cable tension imposes one of the forces to be 8.5 N and the other forces to be larger. In the motion phases and for test 1 (no friction, no cable elasticity), the minimum force varies slightly about 8.5 N depending on the expected motor acceleration. For tests 2 and 3 that take into account the identified friction, we observe that the minimum motor force is  $f_{min} = 2N$  since friction tends to tighten cables and the condition on minimum tension is naturally satisfied. This explains the differences observed during the movement phases between control test 1 where the forces are quite high and control tests 2 and 3 where the forces are almost identical and lower. In stop phases, the forces vary mainly because of the integral correction term. In tests 1 and 2, the forces vary gently. For test 3, since the estimation  $\hat{\alpha}$  depends on the applied forces, the error detected in the control increases with the forces and the integral term increases. We have tried to remedy this problem by using a PD control. By doing so, similar forces as in test 2 have been produced, but the control has become less efficient and tracking errors have increased.

The differences between the  $\hat{\alpha}$  and the values obtained from the video  $\alpha_{IP}$  are shown in Fig. 15. The mean absolute differences are 2.19 deg, 1.51 deg, and 1.09 deg for tests 1, 2, and 3, respectively. In tests 1 and 2, the maximal difference is about 8 deg against 4 deg for test 3. This shows the efficiency of the elastic model. The mean absolute error is lower in test 2 than in test 1. This can be explained by lower applied forces that produce lower elongation and so lower errors on the estimation  $\hat{\alpha}$ .

The differences between the  $\hat{\alpha}$  and the desired values  $\alpha^d$  are shown in Fig. 16. The mean absolute error are 0.52 deg, 0.37 deg, and 0.66 deg for test 1, 2, and 3, respectively. Friction model improves control, with a maximal error twice lower in test 2 than

in test 1. The elastic model, however, disrupts control because this model is used only for the estimation  $\hat{\alpha}$ .

The control law used for test 2 appears as a good compromise: produced forces and control errors are lower than in tests 1 and 3, while the errors between  $\hat{\alpha}$  and the value measured by image processing are acceptable.

Video recording of the second test can be found online.<sup>3</sup>

#### 8 Conclusion

The identification and control of a manipulator built with three tensegrity X-joints in series and remotely actuated by cables has been proposed in this paper. Friction and cable elasticity have been identified with excitation trajectories on a test-bed prototype. Friction stems mainly from the geared motors and is increased by the antagonistic actuation. Pulley friction also plays a role. Both friction effects were identified through the proposed friction model. Regarding cable elasticity, elongation plays a role mainly for the estimation of  $\hat{\alpha}$  that is not directly measured in our test-bed prototype. The cable stiffness is correctly identified but the damping is not identifiable. A model with constant stiffness did not allow us to reconstruct the tension in the cables in an acceptable way, and we proposed a new model with a variable stiffness that increases with the applied forces and seems better suited to our prototype. This identification phase allowed us to build a suitable simulator to understand our experiments. It also allowed us to elaborate several control laws based on the dynamic model of our system. Modeling friction turned out to be of importance since it allows avoiding slack cables without applying large forces. This allows, in turn, limiting energy consumption and cable elongation.

The methodology developed for the test-bed prototype can be applied to manipulators with any number of X-joints. It could be also used for other applications such as cable-driven continuum robots and for cable-driven parallel robots. It is worth noting that antagonistic actuation, which is always present in our study, increases friction since one motor is always counteracting the action of the others. Taking into account friction in the model allows reducing actuation forces and, in turn, cable elongation. Moreover, antagonistic actuation provides additional equations defined by cable redundancy, which makes it possible to reduce errors caused by cable elongation. This explains why, in our context, friction has more significance than cable elasticity for our control. This would not be the case in different applications, such as in suspended cable-driven parallel robots.

In future work, different trajectories corresponding to the main motion types of a bird neck will be tested. In particular, we will experiment explosive motions corresponding to digging into tree trunks. On the other hand, a prototype with ten X-joints is currently under construction. It will be actuated with four cables to stick on the bird's neck architecture. Different cable routing strategies will be compared. Identification and control will be carried out and experiments will be conducted for different typical neck motions.

## Acknowledgment

This work was conducted with the financial support of the French National Research Agency (AVINECK Project ANR-16-CE33-0025). The authors acknowledge their membership to the CNRS GDR2088 BIOMIM research network.

# **Conflict of Interest**

There are no conflicts of interest.

## **Data Availability Statement**

The authors attest that all data for this study are included in the paper. The authors attest that all data for this study are included in the paper.

#### Nomenclature

b = base bar length $\mathbf{c}$  = vector of Coriolis effects affecting the manipulator and geared motors  $\mathbf{g} = \text{gradient of the manipulator}$ potential energy r = pulley radius $\mathbf{v}$  = vector of parameters to identify L = diagonal bar length $\mathbf{M}$  = inertia matrix of the whole manipulator (including geared motors) R = gear ratioW = observation matrix $\mathbf{Z}$  = cable transmission matrix of size  $3 \times 4$  $c_m$  = vector of Coriolis effects affecting the manipulator (without geared motors)  $c_s$  = coefficient used to adjust friction at stop  $\mathbf{f}_{friction} = \text{friction model}$  $f_{\min}$  = minimal actuation force  $r_d = \text{drum radius}$  $t_{\min}$  = minimal tension in cables  $C_x$  = covariance matrix  $I_a$  = motor inertia  $M_m$  = inertia matrix of the manipulator (without geared motors)  $\mathbf{f} = [f_1, f_2, f_3, f_4]_{\mathbf{m}}^T = \text{vector of input forces}$  $\mathbf{k} = [k(1), k(2), k(3), k(4)]^T = \text{vector of cable stiffnesses}$  $\mathbf{k} = [k(1), k(2), k(3), k(4)]' = \text{vector of cable stiffnesses}$   $\mathbf{c}_a = [c_a(1), c_a(2), c_a(3), c_a(4)]^T = \text{vector of cable damping}$   $\mathbf{f}_s = [f_s(1), f_s(2), f_s(3), f_s(4)]^T = \text{vector of friction coefficients}$   $\mathbf{l}_c = [l_{c_1}, l_{c_2}, l_{c_3}, l_{c_4}]^T = \text{vector of cable variable lengths}$   $k_i = \omega^3, k_v = 3\omega, k_p = 3\omega^2 = \text{gains of the PID control}$   $\mathbf{t}_c = [t_{c_1}, t_{c_2}, t_{c_3}, t_{c_4}]^T = \text{vector of cable tensions}$   $\mathbf{x}_c = [x_{c_1}, x_{c_2}, x_{c_3}, x_{c_4}]^T = \text{vector of cable elongations}$   $\alpha = [\alpha_1, \alpha_2, \alpha_3]^T = \text{vector of X-joint orientation}$ angles  $\hat{\alpha}$  = vector of estimation of X-joint orientation  $\alpha_{IP}$  = vector of X-joint orientation angles obtained from video and image processing  $\alpha_d$  = vector of X-joint desired orientation angles  $\psi = [\psi_1, \psi_2, \psi_3]^T$  = vector of angles between the base bar and a diagonal bar of the X-joints  $\theta = [\theta_1, \theta_2, \theta_3, \theta_4]^T = \text{vector of motor angular}$ positions measured at the motor shaft  $\theta^d$  = vector of motor angular positions corresponding to the desired trajectory  $\lambda = [\lambda(1), \lambda(2), \lambda(3), \lambda(4)]^T$  = vector of variable stiffness coefficients  $\sigma_i$  = standard deviation of parameter = vector of residues  $\Gamma^{d}$  = vector of desired torques

<sup>&</sup>lt;sup>3</sup>https://uncloud.univ-nantes.fr/index.php/s/x9PPzXPnJHLGWpm

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