

Towards vision-based control of cable-driven parallel robots

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Abstract—This paper deals with the vision-based control of cable-driven parallel robots. First, a 3D pose visual servoing is proposed, where the end-effector pose is indirectly measured and used for regulation. This method is illustrated and validated on a cable-driven parallel robot prototype. Second, to take into account the dynamics of the platform and using a Cartesian pose and velocity estimator, a vision-based computed torque control is developed and validated in simulation.

I. INTRODUCTION

Parallel mechanisms have a closed-loop structures, with several kinematic chains connecting the base and the mobile platform [1], [2]. Most existing parallel robots are designed with rigid legs which cannot exceed a certain length. A major drawback of such designs is their limited workspace.

Parallel cable-driven robots are a particular type of parallel robots which consist essentially of a moving platform connected in parallel to a base by cables [3]. The length of the cables is adjusted to control the position and orientation of the platform. The cable tensions are related to forces and torques at the moving platform. Parallel cable-driven robots have several advantages over more conventional rigid-link robots. The main one is maybe scalability: Cables with small to very large lengths are easily stored on spools or drums allowing to build parallel cable robots with a workspace of global dimension ranging from a few centimeters to tens of meters or more [4]. Depending on their size and practical realization, they may also be less expensive, easier to build, reconfigure and, when thin cables or wires are used, safer and non intrusive. These useful properties make cable-driven parallel robots good candidates for several applications, e.g., robotic cranes [5] or automated construction systems [6] and human-scale force-feedback haptic systems [7]. Parallel cable-driven robots also have drawbacks such as the unilateral character of actuation (the cables can only pull) which, for some applications, make actuation redundancy compulsory [3].

The control and the modeling of cable-driven parallel robots are complex due to the flexible nature of the cables. In [8], the authors proposed controllers for the cable suspended robot using Lyapunov theory and feedback linearization. In this work, there is no measure of Cartesian position. The pose (position and orientation) of the mobile platform can be estimated through the forward kinematics but the latter is generally difficult to solve and the measures of cable lengths by means of active joint variables may be inaccurate due notably to cable elongations. Another solution is to use ad-



Fig. 1. The LIRMM/Tecnalia ReelAx8 robot

ditional passive cables setting up an independent metrology [9] and whose layout is such that the forward kinematics has closed-loop solutions. This solution increases however the probability of cable collisions and adds constraints to the mechanical design.

Another control strategy was presented in [10]. The proposed method uses a nonlinear feed-forward control laws in cable length coordinates and compute an optimal tension distribution.

In [11], the authors proposed to control a 6-DOF cable robot by means of recursive algorithm to generate reachable domains for motion of the robot end-effector taking into account input constraints. However, continuous tendon forces are not guaranteed. Moreover, due to the complexity of the dynamics of the robot general motion, rotations are not considered in this work. Ref.[12] proposes a real-time algorithm for tendon force calculation, without iterative steps. This approach guarantees the continuity of force distribution.

Several control laws are proposed in [13], [14], [15] using visual servoing techniques [16], [17], [18]. They prove that the use of computer vision in the feedback control loop simplifies the kinematic models by introducing additional sensors into the parallel robots and thus yields easier control.

For the control of cable-driven parallel robots, visual servoing methods can also be a good alternative, since it closes the control loop over the vision sensor. External sensing of the end-effector pose in the feedback signal replaces advantageously the forward kinematic model. The perception models are simpler than the kinematic models and contain less unmodelled physical phenomena. This is especially useful in cable-driven robots because the flexibility of the cable.

Therefore, the contribution of this paper is to present vision-based control of a cable-driven parallel robot. Section II presents the kinematic modeling of the generic

configuration of cable-driven mechanisms. This method is applied to the ReelAx8 parallel robot (Fig.1). Section III-A presents a first method for the vision-based control of cable-driven parallel robots which is 3D pose visual servoing. Then, the vision-based computed torque control approach is presented in Section III-B. Section IV presents an experimental validation of the 3D pose visual servoing as applied to the ReelAx8 robot. Section V shows simulation results of the vision-based computed torque control. Conclusions are finally given in Section VI.

II. VISION-BASED KINEMATICS

In the kinematic modeling the notations given in Table I are used.

- $i = 1..k$ denotes the driving cables, where k is the number of cables.
- Boldface characters denote vectors. Unit vectors are underlined.
- $\mathcal{F}_b = (\mathbf{O}, \underline{\mathbf{x}}_b, \underline{\mathbf{y}}_b, \underline{\mathbf{z}}_b)$, $\mathcal{F}_e = (\mathbf{E}, \underline{\mathbf{x}}_e, \underline{\mathbf{y}}_e, \underline{\mathbf{z}}_e)$ and $\mathcal{F}_c = (\mathbf{O}_c, \underline{\mathbf{x}}_c, \underline{\mathbf{y}}_c, \underline{\mathbf{z}}_c)$ denote the base, end-effector and camera reference frames, respectively.
- ${}^i\mathbf{v}$ is vector \mathbf{v} expressed in \mathcal{F}_i .
- \mathbf{q}_i defines the motorized joint angle i .
- l_i defines the length of cable i .
- ϑ_i is the tension applied by cable i .
- ${}^i\mathbf{T}_j = \begin{pmatrix} {}^i\mathbf{R}_j & {}^i\mathbf{t}_j \\ \mathbf{0} & 1 \end{pmatrix}$ is the homogeneous matrix associated to the rigid transformation from \mathcal{F}_i to \mathcal{F}_j .
- ${}^i\boldsymbol{\tau}_j$ is the cartesian velocity of the origin of \mathcal{F}_j expressed in \mathcal{F}_i .
- \mathbf{X} , $\dot{\mathbf{X}}$ and $\ddot{\mathbf{X}}$ are a representation of the end-effector pose, velocity and acceleration, respectively.
- $\boldsymbol{\Gamma}$ is the vector of motor torques.
- \mathbf{M}^+ is the pseudo-inverse of \mathbf{M} .
- $\hat{\mathbf{M}}$ is the estimation of \mathbf{M} .
- $[\mathbf{a}]_{\times}$ is the cross-product matrix of vector \mathbf{a} .

TABLE I

NOTATIONS USED THROUGHOUT THE PAPER.

ReelAx8 is a 6-DOF cable-driven parallel robot (Fig.1). It has a moving platform (end-effector) connected to a fixed base by 8 driving cables of varying length l_i and cable tension ϑ_i , $i \in 1..8$ ($k = 8$). Each cable (Fig.2) is attached to the moving platform at point \mathbf{B}_i and extends from the base at point \mathbf{A}_i .

We can imagine that ReelAx8 robot as a Gough-Stewart platform [19], [20] which is turned upside down. When the cables are tensed and considered massless and inextensible, the kinematic modeling of Gough-Stewart platforms and of parallel cable-driven robot are very similar.

Fig.2 shows a generic configuration of cable i . The cable-driven parallel robot is equipped with a camera defined by its reference frame \mathcal{R}_c and a pattern defined by frame \mathcal{R}_p .

We consider \mathcal{R}_f a fixed reference frame attached to the base and \mathcal{R}_m a mobile reference frame attached to the end-effector. There are two visual servoing configurations. In the first, (eye-in-hand), the camera attached to the end-effector ($\mathcal{R}_c = \mathcal{R}_m$) observes a pattern fixed with respect to the base

frame ($\mathcal{R}_p = \mathcal{R}_f$). In the second one (eye-to-hand), \mathcal{R}_p is a mobile reference frame ($\mathcal{R}_p = \mathcal{R}_m$) and \mathcal{R}_c is the fixed reference frame ($\mathcal{R}_c = \mathcal{R}_f$).

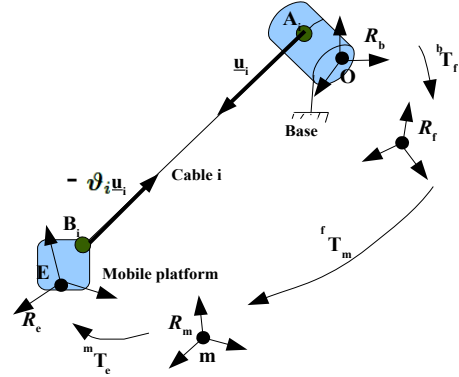


Fig. 2. Generic configuration of cable i

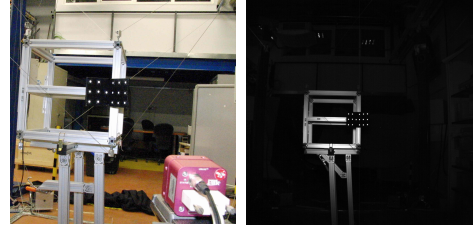


Fig. 3. From left to right: a photograph of ReelAx8 mobile platform and the initial position of the end-effector, seen from the camera.

Introducing $\underline{\mathbf{u}}_i$ the unit vector pointing from \mathbf{A}_i to \mathbf{B}_i (Fig.2), one can write:

$${}^m\overline{\mathbf{A}_i\mathbf{B}_i} = l_i {}^m\underline{\mathbf{u}}_i \quad (1)$$

This equation can be expressed in any Euclidean reference frame. Hence, it can be expressed in the mobile frame \mathcal{R}_m , which simplifies the adequation between the instantaneous inverse kinematic model and the regulated error in the visual servoing task. In the sequel, we consider the ReelAx8 robot equipped with a camera and a pattern in eye-to-hand configuration (Fig.3).

Time differentiating (1), one can obtain:

$$\dot{l}_i = \frac{d({}^m\overline{\mathbf{A}_i\mathbf{B}_i})^T}{dt} {}^m\underline{\mathbf{u}}_i = \frac{d({}^m\mathbf{B}_i - {}^m\mathbf{A}_i)^T}{dt} {}^m\underline{\mathbf{u}}_i \quad (2)$$

where \mathbf{B}_i is time invariant in the mobile frame \mathcal{R}_m , (2) yields:

$$\dot{l}_i = -\frac{d({}^m\mathbf{A}_i)^T}{dt} {}^m\underline{\mathbf{u}}_i = -\frac{d({}^m\mathbf{R}_f^f \mathbf{A}_i + {}^m\mathbf{t}_f)^T}{dt} {}^m\underline{\mathbf{u}}_i \quad (3)$$

Thus, according e.g. to [13], [21], the instantaneous inverse kinematic model of a parallel cable-driven robot, relating the Cartesian velocity ${}^m\boldsymbol{\tau}_m = {}^m\boldsymbol{\tau}_{m/m^*} = ({}^m\mathbf{V}_m \quad {}^m\boldsymbol{\Omega}_m)^T$ of the mobile frame to the time derivative of cable length vector $\dot{\mathbf{l}} = (\dot{l}_1 \quad \dots \quad \dot{l}_k)^T$ is obtained:

$$\dot{\mathbf{i}} = {}^m\mathbf{D}_m^{inv} {}^m\boldsymbol{\tau}_m \quad (4)$$

with

$${}^m\mathbf{D}_m^{inv} = \begin{pmatrix} {}^m\mathbf{u}_1^T & ({}^m\mathbf{B}_1 \times {}^m\mathbf{u}_1)^T \\ \vdots & \vdots \\ {}^m\mathbf{u}_k^T & ({}^m\mathbf{B}_k \times {}^m\mathbf{u}_k)^T \end{pmatrix} \quad (5)$$

where the ${}^m\mathbf{B}_i$ is a constant (calibration) parameter.

The motorized joint angle vector \mathbf{q} of the robot are linearly related to the time derivatives of cable lengths and can be computed as:

$$\dot{\mathbf{q}} = \frac{1}{r_c} \dot{\mathbf{i}} \quad (6)$$

where r_c is the radius of the drums collecting the cables.

Note that the direction of each cable can be written as

$${}^m\mathbf{u}_i = \frac{{}^m\overrightarrow{\mathbf{A}_i\mathbf{B}_i}}{\|{}^m\overrightarrow{\mathbf{A}_i\mathbf{B}_i}\|} = \frac{{}^m\mathbf{B}_i - {}^m\mathbf{A}_i}{\|{}^m\overrightarrow{\mathbf{A}_i\mathbf{B}_i}\|} \quad (7)$$

Being given that ${}^m\mathbf{A}_i = ({}^m\mathbf{R}_f^f \mathbf{A}_i + {}^m\mathbf{t}_f)$, the instantaneous inverse kinematic model (4) depends on the pose ${}^m\mathbf{T}_f$, the rigid transformation between camera and pattern, and on the constant (calibration) parameters ${}^f\mathbf{A}_i$ and ${}^m\mathbf{B}_i$.

III. VISION-BASED CONTROL

A. 3D pose visual servoing

1) Interaction matrix associated to the pose:

Consider \mathcal{R}_m and \mathcal{R}_{m^*} the desired and the current mobile frame positions, respectively. For the control of the robot end-effector pose, in [22], [23], [24], [25] the authors present the following visual primitive \mathbf{s} :

$$\mathbf{s} = \begin{pmatrix} \mathbf{s}_t \\ \mathbf{s}_w \end{pmatrix} \quad (8)$$

In our case, $\mathbf{s}_t = {}^m\mathbf{t}_{m^*}$ is the translation error between \mathcal{R}_m and \mathcal{R}_{m^*} and $\mathbf{s}_w = \mathbf{u}\theta$, where \mathbf{u} is the axis and θ is the angle of the rotation matrix ${}^m\mathcal{R}_{m^*}$. The visual primitive \mathbf{s} is thus $\mathbf{0}$ when the current frame reaches the desired one ($\mathbf{s}^* = \mathbf{0}$).

As shown in [25], [26], we can write:

$$\begin{cases} \dot{\mathbf{s}}_t = -{}^m\mathbf{V}_{m/m^*} + [{}^m\mathbf{t}_{m^*}]_{\times} {}^m\boldsymbol{\Omega}_{m/m^*} \\ \frac{d(\mathbf{u}\theta)}{dt} = \mathbf{L}_w {}^m\boldsymbol{\Omega}_{m^*} = \mathbf{L}_w {}^m\boldsymbol{\Omega}_{m^*/m} \\ = -\mathbf{L}_w {}^m\boldsymbol{\Omega}_{m/m^*} = -\mathbf{L}_w {}^m\boldsymbol{\Omega}_m \end{cases} \quad (9)$$

where

- $\mathbf{L}_w = \mathbf{I}_3 - \frac{\theta}{2}[\mathbf{u}]_{\times} + \left(1 - \frac{\text{sinc}(\frac{\theta}{2})}{\text{sinc}^2(\frac{\theta}{2})}\right)[\mathbf{u}]_{\times}^2$
- $\text{sinc}(\theta) = \frac{\sin(\theta)}{\theta}$

The time derivative of \mathbf{s} is:

$$\dot{\mathbf{s}} = \mathbf{L}_s {}^m\boldsymbol{\tau}_m \quad (10)$$

where ${}^m\boldsymbol{\tau}_m$ is the end-effector Cartesian velocity and \mathbf{L}_s is the interaction matrix expressed as:

$$\mathbf{L}_s = \begin{pmatrix} -\mathbf{I}_3 & [{}^m\mathbf{t}_{m^*}]_{\times} \\ \mathbf{0}_3 & -\mathbf{L}_w \end{pmatrix} \quad (11)$$

2) Control law:

To regulate the error between the current primitive vector \mathbf{s} and the desired one $\mathbf{s}^* = \mathbf{0}$, one can consider the exponential decay $\dot{\mathbf{s}} = -\lambda\mathbf{s}$. The vision-based control (Fig.4) can then be expressed as:

$${}^m\boldsymbol{\tau}_m = -\lambda\widehat{\mathbf{L}_s}^+ \mathbf{s} \quad (12)$$

Let us note that only information from the vision sensor are used to define the relation between \mathbf{s} and the end-effector Cartesian velocity ${}^m\boldsymbol{\tau}_m$ [26].

Note also that the instantaneous inverse kinematic model can be written as:

$$\dot{\mathbf{i}} = {}^m\widehat{\mathbf{D}_m^{inv}} {}^m\boldsymbol{\tau}_m \quad (13)$$

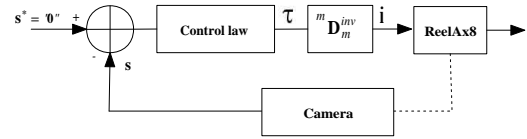


Fig. 4. 3D pose visual servoing of a cable-driven parallel robot

This method allows the execution of a classic and simple robotic task in the Cartesian space.

B. Vision-based computed torque control

In section III-A, we presented a vision-based kinematic control, where the end-effector pose is measured using a camera and used for regulation. This method simplifies the control and replaces joint sensors. However, it does not take into account the dynamics of the robot.

The contribution of this section is to present a vision-based computed torque control for ReelAx8 cable-driven parallel robot (Fig.5).

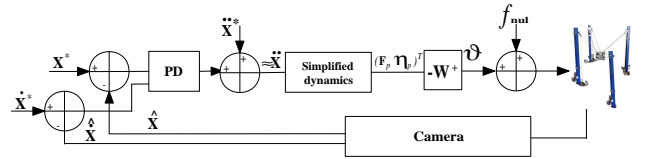


Fig. 5. Vision-based computed torque control of a cable-driven parallel robot

In this section, we consider that the center of mass corresponds to the end-effector origin \mathbf{E} .

1) Simplified dynamics:

In this section, we take into account only the dynamic of the moving platform. The actual ReelAx8 prototype is not a large dimension mechanism. So, the mass of each cable can be neglected to have a simplified dynamics.

The gravitational force applied on the end-effector at the point \mathbf{E} is $m_p\mathbf{g}$. m_p and \mathbf{g} are respectively the mass and the gravity acceleration vector.

Let us denote ${}^f\mathbf{I}_e$ the moment of inertia of the mobile platform about its center of mass expressed in the fixed frame \mathcal{R}_f .

Using the Newton-Euler's laws, one can write the applied and inertia force wrench exerted at point **E** and expressed in the fixed frame \mathcal{R}_f as:

$$\mathcal{T}_{Fe} = \begin{pmatrix} \mathbf{F}_p \\ \boldsymbol{\eta}_p \end{pmatrix}_{\mathcal{R}_f} = \begin{pmatrix} m_p(\mathbf{g} - \mathbf{a}_t) \\ -{}^f\mathbf{I}_e {}^f\dot{\boldsymbol{\Omega}}_e - {}^f\boldsymbol{\Omega}_e \times ({}^f\mathbf{I}_e {}^f\boldsymbol{\Omega}_e) \end{pmatrix} \quad (14)$$

where \mathbf{F}_p is the sum of the applied and inertia forces on the end-effector, $\boldsymbol{\eta}_p$ is the inertia moment and \mathbf{a}_t is the linear acceleration.

2) Equation of motion:

Noting that a cable can only pull the end-effector (Fig.2), one can write the cable tension at point \mathbf{B}_i :

$$\boldsymbol{\vartheta}_i = -\vartheta_i {}^f\mathbf{u}_i \quad (15)$$

where $\vartheta_i = \|\boldsymbol{\vartheta}_i\|$ is tension of cable i which is to be nonnegative.

The wrench applied by cable i at the reference point **E** is:

$$\mathcal{T}_i = \begin{pmatrix} \boldsymbol{\vartheta}_i \\ \boldsymbol{\eta}_{Bi} + {}^f\overrightarrow{\mathbf{EB}_i} \times \boldsymbol{\vartheta}_i \end{pmatrix}_{\mathcal{R}_f} \quad (16)$$

Assuming that the moment of $\boldsymbol{\vartheta}_i$ at point \mathbf{B}_i is $\boldsymbol{\eta}_{Bi} = \mathbf{0}$, we obtain using (15) and (16):

$$\mathcal{T}_i = \vartheta_i W_i \quad (17)$$

where

$$W_i = \begin{pmatrix} -{}^f\mathbf{u}_i \\ -{}^f\overrightarrow{\mathbf{EB}_i} \times {}^f\mathbf{u}_i \end{pmatrix} = \begin{pmatrix} -{}^f\mathbf{u}_i \\ -{}^f\mathbf{R}_e {}^e\mathbf{B}_i \times {}^f\mathbf{u}_i \end{pmatrix} \quad (18)$$

To keep platform in equilibrium, one can write the relation between the applied and inertia forces wrench \mathcal{T}_{Fe} (from (14)) and cable tensions wrenches as:

$$\mathcal{T}_{Fe} + \sum_{i=1}^k (\mathcal{T}_i) = \mathbf{0} \quad (19)$$

It can be also written as:

$$\mathcal{T}_{Fe} + \mathbf{W}\boldsymbol{\vartheta} = \mathbf{0} \quad (20)$$

where $\mathbf{W} = (W_1, \dots, W_k)$, $\boldsymbol{\vartheta} = (\vartheta_1, \dots, \vartheta_k)^T$ and k is the number of cables.

Finally, one can compute the vector of cable tensions as:

$$\boldsymbol{\vartheta} = -\mathbf{W}^+ \mathcal{T}_{Fe} \quad (21)$$

Then, neglecting the inertia of the drums, the motor torques can be computed using:

$$\boldsymbol{\Gamma} = r_c \boldsymbol{\vartheta} \quad (22)$$

where r_c is the reel motor radius of the drums.

As shown in Fig.5, a vector of forces \mathbf{f}_{null} lying in the nullspace of matrix \mathbf{W} can be added to the tension vector computed in (21) in order to try to keep all tensions positive [3] during motion but without changing this latter. The

computation of \mathbf{f}_{null} may be an issue which is beyond the scope of this paper. In the simulation of Section V, \mathbf{f}_{null} is taken equal to the zero vector.

3) Combining measure and computed torque control:

Using a fast exteroceptive measure [15], [27], we can reduce the complexity of control schemes. The proposed computed control torque in Fig.5 needs the estimation of cartesian pose and velocity of the moving platform. They can be measured by vision, which is used directly in the control loop to compensate dynamics in real time.

Given a cartesian error (pose and velocity), we use a proportional and derivative control (PD) to have an estimation of $\ddot{\mathbf{X}}$. Doing so, as shown in section III-B.1, we can easily compute the applied and inertia force wrench. The cable tensions are related to the moving platform dynamics by \mathbf{W} , which can be estimated only by using information from vision (the direction ${}^f\mathbf{u}_i$ and ${}^f\mathbf{R}_e$) and constant calibration parameter ${}^e\mathbf{B}_i$.

IV. EXPERIMENTAL VALIDATION

The 3D pose kinematic visual servoing approach of section III-A has been implemented and applied to ReelAx8 cable-driven parallel robot. An off-the-shelf 'PhotonfocusMV-D1024-TrackCam' camera is used to estimate the pose [27] of the mobile platform.

For technical reasons, the camera is attached to the base frame and a pattern is fixed to the end-effector. The camera is placed in front of the robot so that it can cover a large amount of the workspace (Fig.3).

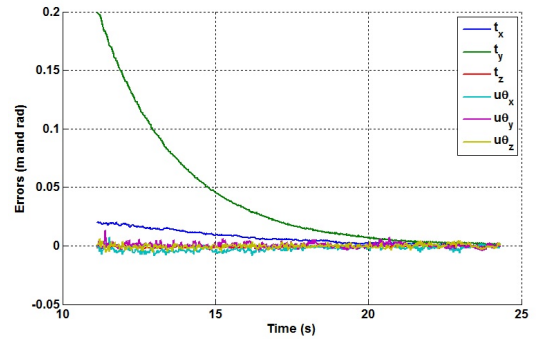


Fig. 6. Evolution of the cartesian error.

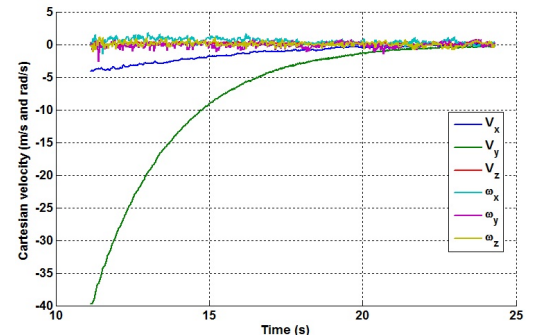


Fig. 7. Evolution of the cartesian velocities.

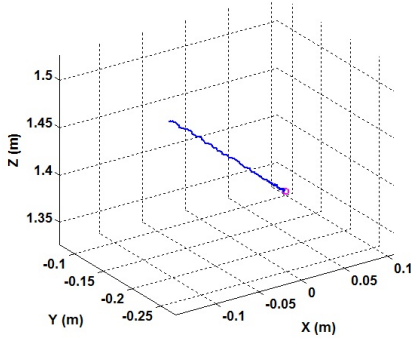


Fig. 8. Evolution of the pattern trajectory.

In an experiment, the ReelAx8 robot is asked to reach the desired position from the initial configuration (Fig.3) ${}^c\mathbf{t}_m = (-0.0181, -0.0744, 1.4248)^T$ (m) as the pattern position and $\theta = 0$ as the rotation angle. The desired position is obtained from the reference position by a translation of -20 mm along the \mathbf{y}_c axis of the camera.

Fig.6 and Fig.7 show that the errors converge to 0 as expected, from an initial cartesian error to a steady state one ($\simeq 1.5$ mm).

Notice that the error curves are exponentials and that the 3D trajectory (Fig.8) of the pattern is linear in camera frame, which confirms the properties of the 3D pose visual servoing.

V. SIMULATION RESULTS

The vision-based computed torque control is validated by means of a ReelAx8 cable-driven parallel robot simulation. All kinematic parameters are expressed in the fixed frame \mathcal{R}_f .

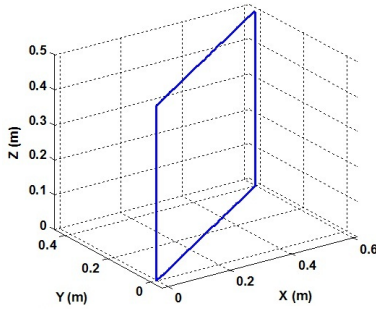


Fig. 9. Evolution of the end-effector trajectory.

In the simulations presented below, we choose the cartesian initial position ${}^b\mathbf{t}_e = (0, 0, 0)^T$ and orientations $\alpha = \beta = \gamma = 0$.

Fig.9 and Fig.10 show the reference rectangular motion. It is composed of translations along the three axis \mathbf{x}_b , \mathbf{y}_b and \mathbf{z}_b . The last part of the reference trajectory (Fig.10) is composed of rotations around the three axis \mathbf{x}_b , \mathbf{y}_b and \mathbf{z}_b .

To show the robustness of the approach, we added noise to the estimation of the cartesian pose ${}^b\mathbf{T}_e$ and velocity. Consequently, we define at each time a sample random noise for translation and operational velocity, with maximal value

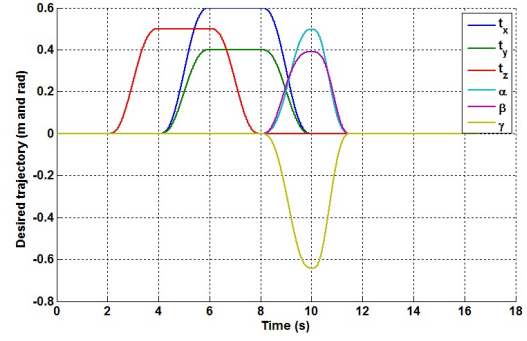


Fig. 10. Desired position and orientation of the end-effector.

of 0.2 mm and 2 mm/s, respectively. We also added noise to the rotation angle and angular velocity, with maximal amplitude of 0.01° and of $0.1^\circ/s$, respectively.

Fig.11, Fig.12 and Fig.13 show a potentially good robustness. Table II presents mean values and standard variances (deviations) of the error vector norm.

	Mean value	Standard variance
Translation error (mm)	0.4	0.1
Orientation error ($^\circ$)	0.0225	0.0038
Operational velocity error (mm/s)	1.1	3.6
Orientation velocity error ($^\circ/s$)	0.0616	0.2069

TABLE II

MEAN VALUES AND STANDARD VARIANCES OF CARTESIAN ERRORS

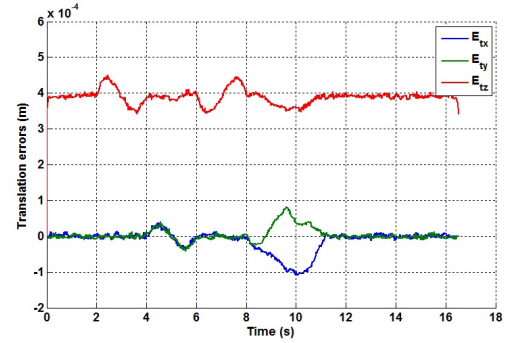


Fig. 11. Evolution of the translation errors.

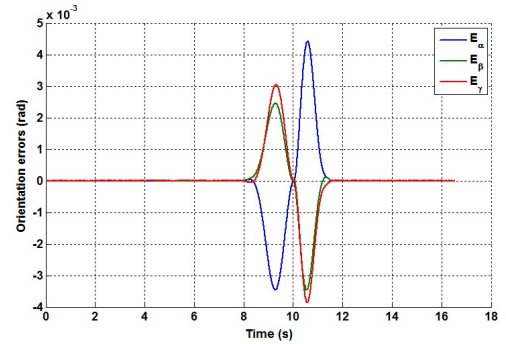


Fig. 12. Evolution of the orientation errors.

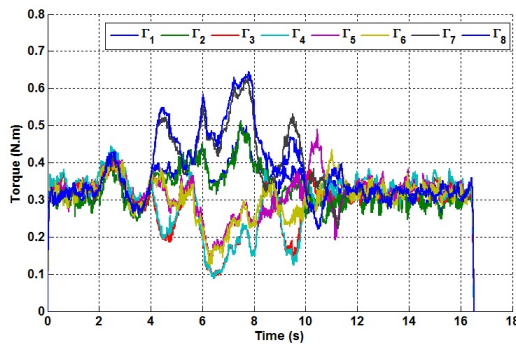


Fig. 13. Evolution of motor torques.

VI. CONCLUSION

This paper proposed methods to control a cable-driven parallel robot using a visual target and a calibrated camera. The use of computer vision in the feedback simplifies the kinematic modeling by introducing additional sensors into the robot and thus yields easier control.

Using a 3D pose kinematic visual servoing method, the end-effector pose is used for regulation. This method validated on a cable robot prototype, ReelAx8, robot differs from standard control in the joint space by the fact that the forward kinematic model is never used in the control.

It has to be noticed that the pose is estimated in the task space which is the Cartesian space of the robot. Consequently, we wanted to fuse a 3D pose and velocity visual tracking [27] and a dynamic control of cable-driven parallel robots.

Doing so, we have introduced a second approach: a vision-based computed torque control. The presented method was validated on a ReelAx8 robot simulator. It is performed with noise added to show the validity of the approach.

The next step in this work is to implement the vision-based computed torque control on the ReelAx8 robot and to extend the method to large dimension cable-driven parallel robots.

VII. ACKNOWLEDGMENT

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