Aerial Co-Manipulation with Cables: The Role of Internal Force for Equilibria, Stability, and Passivity
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Marco Tognon\textsuperscript{1}, Chiara Gabellieri\textsuperscript{1,2,†}, Lucia Pallottino\textsuperscript{2}, and Antonio Franchi\textsuperscript{1}

Abstract—This paper considers the cooperative manipulation of a cable-suspended load with two generic aerial robots without the need of explicit communication. The role of the internal force for the asymptotic stability of the beam position-and-attitude equilibria is analyzed in depth. Using a nonlinear Lyapunov-based approach, we prove that if a non-zero internal force is chosen, then the asymptotic stabilization of any desired beam attitude can be achieved with a decentralized and communication-less master-slave admittance controller. If, conversely, a zero internal force is chosen, as done in the majority of the state-of-the-art algorithms, the attitude of the beam is not controllable without communication. Furthermore, we formally proof the output-strictly passivity of the system with respect to an energy-like storage function and a certain input-output pair. This proves the stability and the robustness of the method during motion and in non-ideal conditions. The theoretical findings are validated through extensive simulations.

I. INTRODUCTION

Over the last decade UAVs (Unmanned Aerial Vehicles) have risen the interest of a larger and larger audience for their wide application domain. Recently, aerial physical interaction, using aerial manipulators [1], [2] or exploiting physical links as cables [3], has become a very popular topic. One interesting and applicative problem is the aerial manipulation of large objects, for which cooperative approaches are usually applied because they allow to overcome the limited payload of a single platform, thus lifting larger and heavier loads [4].

Many works targeted this problem proposing different methods and solutions. In [5], [6] cooperative aerial transportation of a rigid and an elastic object is considered, respectively. In [7] the use of multiple flying arms is exploited to address the problem. Aerial manipulation via cables is another interesting solution to the problem since it can reduce the couplings between the load and the robot attitude dynamics. Examples of cooperative aerial manipulation using cables are studied in [8]–[10]. All these examples rely on a centralized control. Instead, a decentralized algorithm, as in [11], is more robust and scalable with respect to (w.r.t.) the number of robots.

However, the major bottleneck in decentralized algorithms is the explicit communication. Communication delays and packet losses can affect the performance and even the stability of the systems. Limiting the need for explicit communication allows to reduce the complexity as well. In [12] the authors proposed one of the first decentralized leader-follower algorithm without explicit communication, for objects transportation performed by mobile ground robots. Aerial cooperative transportation by two robots without explicit communication has been addressed also in [13] for a cable-suspended beam-like load, and a leader-follower paradigm has been proposed. Here the leader follows an external position reference, while the horizontal position of the follower is controlled with an admittance filter, trying to keep the cable always vertical (zero internal force). A similar approach has been proposed in [14] but relying on a visual feedback. However, those methods do not deal with the load pose control and do not provide a formal stability proof.

For the same system composed by two aerial robots carrying a cable suspended beam-like load (see Fig. 1 for a schematic representation), we propose a decentralized algorithm relying only on implicit communication. Our algorithm uses a master-slave architecture with an admittance filter on both robots (not only on the slave as in the related state of the art), to make the overall system compliant/robust to external disturbances.

One of our main contributions is the constructive and intuitive method to choose the controller input to stabilize the load at a desired pose. The control of both position and orientation turns the simpler transportation task found in the state of the art in a full-manipulation one.

We show that those inputs are parametrized by the internal force of the load that plays a crucial role in the equilibria stability. Differently from the state of the art algorithms, which are not formally guaranteed to converge, we also provide a formal proof of the stability through Lyapunov’s
direct method. Furthermore, we prove that the controlled system is output-strictly passive w.r.t. a relevant input-output pair. This provides a bound for the energy variations during the manipulation and an index of robustness of the method.

In Sec. II we derive the model. In Sec. III we present the control strategy and the equilibria of the system. Their stability is discussed in Sec. IV. In Sec. V we prove the passivity and stability of transportation. Simulation results and conclusive discussions are presented in Sec. VI and VII, respectively.

II. SYSTEM MODELING

The considered system and its major variables are shown in Fig. 1. The beam-like load is modeled as a rigid body with mass $m_l \in \mathbb{R}_{>0}$ and a positive definite inertia matrix $J_l \in \mathbb{R}^{3 \times 3}$. We define the frame $F_L = \{O_l, x_l, y_l, z_l\}$ rigidly attached to it, where $O_l$ is the load center of mass (CoM).

Then, we define an inertial frame $F_W = \{O_w, x_w, y_w, z_w\}$ with $z_w$ oriented in the opposite direction to the gravity vector. The configuration of the load is then described by the position of $O_l$ and orientation of $F_L$ with respect to $F_W$, i.e., by the vector $p_l \in \mathbb{R}^3$ and the rotation matrix $R_l \in SO(3)$, respectively. Its dynamics is given by the Newton-Euler equations

$$m_l \ddot{p}_l = -m_l g e_3 + f_e,$$
$$R_L = S(\omega_l)R_l,$$
$$J_L \dot{\omega}_l = -S(\omega_l)J_l \omega_l + \tau - \omega_l^T B_l \omega_l,$$

where, $\omega_l \in \mathbb{R}^3$ is the angular velocity of $F_L$ w.r.t. $F_W$ expressed in $F_L$. $S(\cdot)$ is the operator such that $S(x) = x \times y$. $g$ is the gravitational constant, $e_3$ is the canonical unit vector with a 1 in the $i$-th entry, $f_e$ and $\tau \in \mathbb{R}^3$ are the sum of external forces and moments acting on the load, respectively. The positive definite matrix $B_l \in \mathbb{R}^{3 \times 3}$ is a damping factor modeling the energy dissipation phenomena.

The load is transported by two aerial robots by means of two cables, one for each robot. We denote with $A_i$ the attachment point of the $i$-th cable to the $i$-th robot, with $i = 1, 2$, and we define the frame $F_{Ri} = \{A_i, x_{Ri}, y_{Ri}, z_{Ri}\}$ rigidly attached to the robot and centered in the attachment point. The $i$-th robot configuration is described by the position of $A_i$ and orientation of $F_{Ri}$ w.r.t. $F_W$, denoted by the vector $p_{Ri} \in \mathbb{R}^3$, and the rotation matrix $R_{Ri} \in SO(3)$, respectively. We assume that a position controller is applied to the aerial robot, able to track any $C^2$ trajectory with negligible error in the domain of interest, independently from external disturbances. Indeed, with the recent robust controllers (as the one in [15] for both unidirectional- and multidirectional-thrust vehicles) and disturbance observers for aerial vehicles, one can obtain very precise motions, even in the presence of external disturbances. However, the proposed control method results particularly robust to non-idealities, thanks to its passivity nature (see Sec. V). As a consequence, in real applications, a precise tracking is actually not needed for the stability, but only to achieve perfect performance.

The closed loop translational dynamics of the robot subject to the position controller is then assumed as the one of a double integrator: $p_{Ri} = u_{Ri}$, where $u_{Ri}$ is a virtual input to be designed. If we consider a multidirectional-thrust platform capable of controlling both position and orientation independently [16], the double integrator is an exact model of the closed loop system apart from modeling errors. In the case of underactuated unidirectional-thrust vehicle, the double integrator is instead a very good approximation. Indeed the rotational dynamics is totally decoupled from the translational one and it is much faster than the latter, allowing to apply the time-scale separation principle. At this stage it might seem that the platform is ‘infinitely stiff’ w.r.t. the force produced by the cable. However, we shall re-introduce a compliant behavior by suitably designing the input $u_{Ri}$.

The other end of the $i$-th cable is attached to the load at the anchoring point $B_i$ described by the vector $^t b_i \in \mathbb{R}^3$ denoting its position with respect to $F_L$. The position of $B_i$ w.r.t. $F_W$ is then given by $b_i = p_{Ri} + R_{Ri} ^t b_i$. To simplify the discussion we assume, without loss of generality, that $L b_1 = ([^t b_1] \ 0 \ 0)^T$.

Assumption 1. The two anchoring points are placed such that the load CoM coincides with their middle point, i.e., $^t b_1 = -^t b_2$. This assumption is rather easy to meet in practice.

We model the $i$-th cable as a unilateral spring along its principal direction, characterized by a constant elastic coefficient $k_i \in \mathbb{R}_{>0}$, a constant nominal length denoted by $l_0$ and a negligible mass and inertia w.r.t. the ones of the robots and of the load. The attitude of the cable is described by the normalized vector, $n_i = l_i / \|l_i\|$, where $l_i = p_{Ri} - b_i$. Given a certain elongation $\|l_i\|$ of the cable, the latter produces a force acting on the load at $B_i$ equal to:

$$f_i = t_i n_i, \quad t_i = \begin{cases} k_i(\|l_i\| - l_0) & \text{if } \|l_i\| - l_0 > 0 \\ 0 & \text{otherwise} \end{cases}. \tag{1}$$

$t_i \in \mathbb{R}_{>0}$ denotes the tension along the cable and it is given by the simplified Hooke’s law. As usually done in the related literature, we assume that the controller and the gravity force always maintain the cables taut, at least in the domain of interest. The force produced at the other hand of the cable, namely on the $i$-th robot at $A_i$, is equal to $-f_i$.

Considering the forces that robots and load exchange by means of the cables, the dynamics of the full system is:

$$\dot{v}_R = u_R,$$
$$\dot{v}_L = M_L^{-1} (-c_L(v_L) - g_L + G(q_L) f),$$

where $q_R = [p_{R1} \ p_{R2}]^T$, $q_L = [p_L \ R_L]^T$, $v_R = [p_{R1} \ p_{R2}]^T$, $v_L = [p_L \ \omega_l]^T$, $u_R = [u_{R1} \ u_{R2}]^T$, $f = [f_1 \ f_2]^T$ where $f_i$ is given in (1), and is a function of the state, $M_l = \text{diag}(mj^2 I_3)$ and $I_3 \in \mathbb{R}^{3 \times 3}$ the identity matrix, $g_L = [-m_l g e_3 \ 0]^T$, $c_L = [0 \ S(\omega_l)J_l \omega_l - \omega_l^T B_l \omega_l]^T$ and $G = \begin{bmatrix} I_3 & I_3 \\ S(^t b_1) R_L & S(^t b_2) R_L \end{bmatrix}$.

We remark that the two dynamics in (2) are coupled together by the cable forces in (1).
Control problem

In this work we aim to: i) stabilize the load at a desired configuration, \( \bar{q}_L = (\bar{p}_L, \bar{R}_L) \); ii) preserve the stability of the load during its transportation.

Assuming a perfect knowledge of the system dynamic model, and a perfect state estimation, one could use a model-based observer as done in \cite{13}, measured by an on-board force sensor or estimated by a standard on-board sensors, while the second can be directly required explicit communication. Indeed it requires only local virtual damping, and the stiffness of a virtual spring attached to the robots.

III. CONTROL DESIGN AND EQUILIBRIA

To achieve the previous control objectives we propose the use of an admittance filter for both robots, i.e., setting:

\[
u_i = M_{A_i}^{-1} (-B_A \dot{p}_{R_i} - K_A p_{R_i} - f_i + \pi_{A_i}),
\]

where the tree positive definite symmetric matrices \( M_{A_i}, B_A, K_A \in \mathbb{R}^{3 \times 3} \) are the virtual inertia of the robot, the virtual damping, and the stiffness of a virtual spring attached to the robot, and \( \pi_{A_i} \in \mathbb{R}^3 \) is an additional input (see Fig. 2 for a schematic representation). Notice that (3) does not require explicit communication. Indeed it requires only local information, i.e., the state of the robot \( (\dot{p}_{R_i}, \dot{p}_{R_i}) \), and the force applied by the cable \( f_i \). The first can be retrieved with standard on-board sensors, while the second can be directly measured by an on-board force sensor or estimated by a sufficiently precise model-based observer as done in \cite{13, 16}.

Combining equations (2) and (3) we can write the closed loop system dynamics as \( \dot{v} = m(q, v, \pi_A) \) where

\[
m(q, v, \pi_A) = \begin{bmatrix} M_A^{-1} (-B_A \dot{p}_{R} - K_A p_{R} - f + \pi_A) \\ M_L^{-1} (-c_L (v_L) - g_L + Gf) \end{bmatrix},
\]

with \( q = (q_R, q_L), v = [v_R, v_L]^\top \) and \( \pi_A = [\pi_{A_1}, \pi_{A_2}]^\top \). Furthermore \( M_A = \text{diag}(M_{A_1}, M_{A_2}), B_A = \text{diag}(B_{A_1}, B_{A_2}) \) and \( K_A = \text{diag}(K_{A_1}, K_{A_2}) \). In order to coordinate the motions of the robots in a decentralized way we propose a master-slave approach. Only one robot, namely the designated master, will have an active control of the system. Choosing robot 1 as master and robot 2 as slave we set \( K_{A_1} \neq 0, K_{A_2} = 0 \).

We say that \( q \) is an equilibrium configuration if \( \exists \pi_A \) s.t. \( 0 = m(q, 0, \pi_A) \), i.e. if the corresponding zero-velocity state \( (q, 0) \) is a forced equilibrium for the system (4) for a certain forcing input \( \pi_A \). We say that an equilibrium configuration \( q \) is stable, unstable, or asymptotically stable if \( (q, 0) \) is stable, unstable, or asymptotically stable, respectively.

In the following we shall prove that for any desired load configuration \( \bar{q}_L \) there exists a set \( \Pi_A(\bar{q}_L) \subset \mathbb{R}^6 \) such that for any \( \pi_A \in \Pi_A(\bar{q}_L) \) one can compute a \( \bar{q}_R \), depending on \( \bar{q}_L \) and \( \pi_A \), that makes \( \bar{q} = (\bar{q}_L, \bar{q}_R) \) an asymptotically stable equilibrium with \( \pi_A \) as forcing input. As we shall see, a key role in all the following analyses is played by the load internal force, defined as

\[
t_L := \frac{1}{2} f^\top [I_3 - I_3]^\top R_L e_1 =: \frac{1}{2} f^\top r_L,
\]

where \( r_L = [I_3 - I_3]^\top R_L e_1 \). We have that if \( t_L > 0 \) the internal force is a tension (the work of the internal force is positive if the distance between the anchoring points increases) while if \( t_L < 0 \) the internal force is a compression (vice versa, the work is positive if the distance decreases).

A. Equilibrium Configurations of the Closed Loop System

We firstly carefully analyze the relation between equilibrium configurations, from now on simply called equilibria, and the forcing input \( \pi_A \). In particular, we shall study: i) equilibria inverse problem: which is the set of inputs (and corresponding robot positions) that equilibriates a desired \( \bar{q}_L \) (Theorem 1); ii) equilibria direct problem: which is the set of equilibria if \( \pi_A \), chosen in the aforementioned set, is applied to the system (Theorem 2). A schematic representation of the results described in the theorems is given in Fig. 3.

**Theorem 1** (equilibria inverse problem). Consider the closed loop system (4) and assume that the load is at a given desired configuration \( q_L = \bar{q}_L = (\bar{p}_L, \bar{R}_L) \). For each internal force \( t_L \in \mathbb{R} \), there exists an unique constant value for the forcing input \( \pi_A = \bar{\pi}_A \) (and an unique position of the robots \( q_R = \bar{q}_R \)) such that \( \bar{q} = (\bar{q}_L, \bar{q}_R) \) is an equilibrium of the system.

In particular \( \bar{\pi}_A \) and \( \bar{q}_R = [\bar{p}_R^\top, \bar{R}_R^\top]^\top \) are given by

\[
\bar{\pi}_A(q_L, t_L) = K_A \bar{q}_R + \bar{f}(\bar{q}_L, t_L)
\]

\[
\bar{p}_R(q_L, t_L) = \bar{p}_L + \bar{R}_L b_i + \left( \frac{||\bar{f}_i||}{k_i} + \zeta_0 \right) \frac{\bar{f}_i}{||\bar{f}_i||},
\]

for \( i = 1, 2 \), where

\[
\bar{f}(\bar{q}_L, t_L) = \begin{bmatrix} \bar{f}_1 \\ \bar{f}_2 \end{bmatrix} = \frac{m g}{2} \begin{bmatrix} I_3 \\ I_3 \end{bmatrix} e_3 + t_L \begin{bmatrix} I_3 \\ -I_3 \end{bmatrix} \bar{R}_L e_1.
\]
Theorem 2. The desired load configuration \( \hat{q}_L \) can be equilibrated if there exists at least a \( \hat{q}_R \) and a \( \pi_A \) such that:

\[
m(\hat{q}, 0, \pi_A) = 0.\tag{9}
\]

Consider the last six rows of (9). We must find the \( f \) solving

\[
Gf = g_L.\tag{10}
\]

\( G \) is not invertible since \( \text{rank}(G) = 5 \), thus we have to verify that a solution for (10) exists. Expanding (10) we obtain

\[
f_1 + f_2 = -m_L e_3
\]

(11)

\[
S(\hat{b}_1) \hat{R}_1 f_1 + S(\hat{b}_2) \hat{R}_2 f_2 = 0.\tag{12}
\]

Then, substituting in (12) the \( f_1 \) obtained from (11) we have

\[
2S(\hat{b}_1) \hat{R}_1 f_1 = -S(\hat{b}_1) \hat{R}_1 m_L e_3, \text{ for which } f_2 = m_L e_3/2 \text{ is always a solution. Therefore, all the solutions of (10) can be written as}
\]

\[
\tilde{f} = G^\dagger g_L + r_L f_L,\tag{13}
\]

where \( G^\dagger = 1/2[I_3 I_3]^T \) is the pseudo inverse of \( G \), \( r_L \in \mathbb{R}^6 \) is a vector in \( \text{null}(G) \), and \( f_L \in \mathbb{R} \) is an arbitrary number.

We computed \( f_L = [f_1^T f_2^T]^T \) from (11) and (12) imposing the right hand side equal to zero. From (11) \( f_2 = -f_1 \), and replacing it into (12) we obtain

\[
S(\hat{b}_1) \hat{R}_1 f_1 = 0 \text{ which is verified if } f_1 = 0. \text{ Finally we obtain}
\]

\[
r_L = [I_3 - I_3]^T \hat{R}_1, \text{ as in the definition (5)}.
\]

Equation (13) can be then rewritten as (8). The expression of \( \bar{p}_R \) in (7) is computed using (1) and the kinematics of the system. Notice that (7) is singular when \( \bar{f}_i = 0 \) for some \( i \). However this can always be avoided properly choosing \( t_L \).

Lastly, from the first six rows of (9) we have that \( \hat{q}_L \) is equilibrated if \( \pi_A = \bar{\pi}_A \), where \( \bar{\pi}_A \) is defined as in (6).

Remark 1. Based on Theorem 1 we can define a set \( \Pi_{\bar{L}}(\hat{q}_L) = \{ \pi_A \in \mathbb{R}^6 : \pi_A = \bar{\pi}_A(\hat{q}_L, t_L) \text{ for } t_L \in \mathbb{R} \} \) which has dimension 1, since it is parametrized by the scalar \( t_L \in \mathbb{R} \).

Remark 2. Given a desired load configuration \( \hat{q}_L \) to equilibrate, Theorem 1 and its constructive proofs, give an intuitive method for choosing the forcing input \( \pi_A \). In particular one has to choose only the value of the internal force \( t_L \).

Once \( t_L \) is chosen and the input \( \pi_A = \bar{\pi}_A(t_L, \hat{q}_L) \) is applied to the system, it is not in general granted that \( \hat{q}_L, \bar{q}_R \) is the only equilibrium of (4), i.e., the equilibria direct problem may have multiple solutions.

Theorem 2 (equilibria direct problem). Given \( t_L \in \mathbb{R} \) and the corresponding \( \bar{\pi}_A \in \Pi_{\bar{L}}(\hat{q}_L) \) computed as in (6), the equilibria of the system (4), when the input \( \pi_A = \bar{\pi}_A(t_L, \hat{q}_L) \) is applied, are all and only the ones described by the following conditions

\[
t_L \hat{R}_L e_1 \times \hat{R}_L e_1 = 0
\]

(14)

\[
p_{R1} = \bar{p}_{R1}
\]

\[
p_L = p_{R1} - t_L \hat{b}_1
\]

\[
= p_L + (\hat{R}_L - R_L) \hat{b}_1
\]

(14)

\[
p_{R2} = p_L + t_L \hat{b}_2 + (\frac{\| \hat{f}_2 \|}{k_2} + l_0) \hat{f}_2
\]

(14)
Theorem 3. Let us consider a desired load configuration $\bar{q}_L$. For the system (4) let the constant forcing input $\pi_A$ be chosen in $\Pi_A(\bar{q}_L)$ corresponding to a certain internal force $t_L$. Then $x$ belonging to:

- $\mathcal{X}^+(t_L, \bar{q}_L)$ is locally asymptotically stable if $t_L > 0$;
- $\mathcal{X}^-(t_L, \bar{q}_L)$ is unstable if $t_L > 0$;
- $\mathcal{X}^+(\bar{q}_L, \bar{q}_L)$ is locally asymptotically stable;
- $\mathcal{X}^-(\bar{q}_L, \bar{q}_L)$ is unstable if $t_L < 0$.

Proof. Let us consider the following Lyapunov candidate:

$$V(x) = \frac{1}{2} (v_r^T M_r v_r + e_r^T K_a e_r + v_L^T M_L v_L + k_1 (||l|| - l_0)^2 + k_2 (||x|| - l_0)^2) - l_1^T f_1 + l_2^T f_2 + t_L (1 - (R_L e_1)\dagger R_L e_1) + V_0,$$

where $V_0 \in \mathbb{R}_{>0}$ and $e_r = \bar{p}_{R1} - p_{R1}$. For an optimum choice of $V_0$, $V(x)$ is a positive definite, continuously differentiable function in the domain of interest for which we have that $x_{\min} = \arg \min_x V(x)$ such that $x_{\min} \in \mathcal{X}(0, \bar{q}_L)$. The proof is provided in technical detail in the multimedia materials. In particular, if $t_L \geq 0$, we can choose the term $V_0$ such that $V(x) \geq 0$ for all $x \in \mathcal{X}(0, \bar{q}_L)$ and $x \in \mathcal{X}^+(t_L, \bar{q}_L)$ for $t_L > 0$. Notice that $V(x) = 0$ for all $x \in \mathcal{X}(0, \bar{q}_L)$ and $x \in \mathcal{X}^+(t_L, \bar{q}_L)$ for $t_L > 0$.

Computing the time derivative of (18) and replacing (4), (1) and (8) we obtain:

$$\dot{V} = -v_r^T B_1 v_r - \omega_1^T B_1 \omega_L L$$

is clearly negative definite. In particular, $\dot{V}(x) = 0$ for all $x \in \mathcal{X}(x : v_r = 0, \omega_L = 0)$.

Since $V$ is only negative definite, to prove the asymptotic stability we rely on the LaSalle’s invariance principle [17]. Let us define a positively invariant set $\Omega_{\alpha} = \{x : V(x) < \alpha \text{ with } \alpha \in \mathbb{R}_{>0}\}$. By construction $\Omega_{\alpha}$ is compact since (18) is radially unbounded and $\Omega_{\alpha}$ is compact for $\Omega_{\alpha} = \mathcal{X}(t_L, \bar{q}_L)$, and $\Omega_{\alpha} = \mathcal{X}^+(t_L, \bar{q}_L)$, for $t_L = 0$ and $t_L > 0$, respectively, are both compact sets. Then we need to find the largest invariant set $M$ in $\mathcal{X}$ such that $V(x) = 0 \Rightarrow \dot{V}(x) \leq 0$ and $\omega_L(t) = 0 \Rightarrow \omega_L(t) = 0$ for all $t \in \mathbb{R}_{>0}$. Therefore $x$ has to be an equilibrium, and from Theorem 2 we have that $V(x(t)) = 0 \Rightarrow x(t) \in \mathcal{X}(t_L, \bar{q}_L)$. Thus we obtain $M = \Omega_{\alpha} \cap \mathcal{X}(t_L, \bar{q}_L)$.

For $t_L > 0$, it is easy to see that for a sufficiently small $\alpha$, $\mathcal{X}^+(t_L, \bar{q}_L) \subseteq \Omega_{\alpha}$, but $\mathcal{X}^-(t_L, \bar{q}_L) \cap \Omega_{\alpha} = \emptyset$. Therefore, $\mathcal{X}^+(t_L, \bar{q}_L)$ is locally asymptotically stable for the LaSalle’s invariance principle.

On the other hand, for $t_L = 0$ we have that $\mathcal{X}^+(t_L, \bar{q}_L) \subseteq \Omega_{\alpha}$ for every sufficiently small $\alpha$. Therefore $\mathcal{X}^+(t_L, \bar{q}_L)$ and, as before, we can conclude that $\mathcal{X}^+(t_L, \bar{q}_L)$ is locally asymptotically stable for the LaSalle’s invariance principle.

V. PASSIVITY AND STABILITY OF MANIPULATION

Theorem 3 characterizes the stability of all the possible static equilibria given a certain constant forcing input. In particular, it shows that one has to choose $t_L > 0$ and $\pi_A \in \Pi_A(\bar{q}_L)$ to let the system asymptotically converge to a desired load configuration. On the contrary, one must avoid $t_L = 0$ because the control of the load attitude and its position is not possible. Notice that this last case is the most used in the literature, where the attempt is made to keep the cables always vertical, i.e., with no internal forces.

Let us now show how one can exploit the input $\pi_{A1}$ in order to move the load between two distinct positions. From (6)-(8) and from the fact that $K_B = 0$, it descends that only $\pi_{A1}$, in $\pi_A = [\pi_{A1} \pi_{A2}]^T$, actually depends on the desired load position $\bar{p}_L$. This makes robot 1 able to steer alone the load position without communicating with robot 2. This is
done by first plugging a new desired position $\tilde{p}_L$ in (6) thus computing a new $\tilde{p}'_{RL}$, and then plugging $\tilde{p}'_{RL}$ in (7) in order to compute the new constant forcing input $\tilde{\pi}'_{A1}$. However, one may want to minimize the transient phases generated by a piecewise constant forcing input. It is sufficient to design $\pi_{A1}$ as

$$\pi_{A1}(t) = \pi_{A1} + u_{A1}(t),$$

(19)

where $u_{A1}(t)$ is a smooth function such that $\pi_{A1}(0) = \tilde{\pi}_{A1}$ and $\pi_{A1}(t_f) = \tilde{\pi}'_{A1}$ for $t_f \in \mathbb{R}^{+}$.

To ensure that the system remains stable when the input is time-varying, we shall prove that the system is output-strictly passive w.r.t. the input-output pair $(u, y) = (u_A, v_R)$.

**Theorem 4.** If $\pi_A$ is defined as in (19) for a certain $\tilde{q}$ and $\tilde{q}'$ with $t_L \geq 0$, then system (4) is output-strictly passive w.r.t. the storage function (18) and the input-output pair $(u, y) = (u_A, v_R)$.

**Proof.** In the proof of Theorem 3 we already shown that (18) is a continuously differentiable positive definite function for $t_L \geq 0$, properly choosing $V_0$. Furthermore, replacing (19) into (3), and differentiating (18) we obtain

$$\dot{V} = -v_R \top B_A v_R + v_R \top u - \omega_L \top B_L \omega_L \leq u \top y - y \top B_A y = u \top y - y \top \Phi(y),$$

with $y \top \Phi(y) > 0 \forall y \neq 0$. Therefore, system (4) is output-strictly passive [17].

Thanks to the passivity of the system we can say that for a bounded input provided to the master, the energy of the system remains bounded too, and in particular it stabilizes to a new constant value as soon as $u_{A1}$ becomes constant again. This means that while moving the master, the overall state of the system will remain bounded, and will converge to another specific equilibrium configuration when the master input becomes constant. Furthermore, it is well known that passivity is a robust property, especially w.r.t. model uncertainties. In particular, choosing $\pi_A \in \Pi_A(\tilde{q}_L)$ for a given $\tilde{q}$, the system remains asymptotically stable even in the presence of some parameter uncertainties, but it will converge to a $\tilde{q}$ that is slightly different from $\tilde{q}$.

**Remark 4.** Once the desired load pose is decided and the value of $t_L$ is chosen, one can compute the control input $\pi_A$ and send it to the robots. Afterwards, if $t_L > 0$ the robots will steer the load to the desired configuration preserving the stability and without the need of sending data to each other. The cooperative task is performed exploiting the implicit communication through the forces that the robots exchange and feel from the cables and the object.

**VI. NUMERICAL VALIDATION**

In this section we shall describe the results of several numerical simulations validating the proposed method and all the presented theoretical concepts and results.

For the simulation we considered a quadrotor-like vehicle with its proper nonlinear dynamics together with a geometric position controller, even though, our method can be applied to more general flying vehicles. System and control parameters are reported in Tab. I. Notice the smaller apparent inertia of the slave, chosen to make it more sensitive to external forces.

Let us consider the desired equilibrium $\dot{\tilde{q}} = (\tilde{p}_L, \tilde{R}_L)$, whose value are in Tab. I, where $(\phi, \theta, \psi)$ are the Euler angles that parametrize $\tilde{R}_L$. We performed several simulations with $\pi_A \in \Pi_A(\tilde{q}_L)$ computed as in (6) for the cases: 1) $t_{L,1} = 1.5$ [N] > 0, 2) $t_{L,2} = 0$ [N], 3) $t_{L,3} = -1$ [N] < 0.

To test the stability of the equilibria, we initialized the system in different initial configurations and we let it evolve. Figure 6 shows the position and orientation error for the three $t_L$ and several different initial conditions. 1) For $t_L = t_{L,1}$, the system always converges to a state belonging to $\mathcal{X}^{++}(t_L, \tilde{q}_L)$, independently from the initial state, validating the asymptotic stability of $\mathcal{X}^{++}(t_L, \tilde{q}_L)$ when $t_L > 0$. 2) For $t_{L,2}$, the system final state belongs to $\mathcal{X}(0, \tilde{q}_L)$. The particular final attitude of the load depends on the initial state. 3) For $t_{L,3}$, the system never converges to $\mathcal{X}^{++}(t_L, \tilde{q}_L)$ even with a very close initial configuration. This is due to the instability of $\mathcal{X}^{++}(t_L, \tilde{q}_L)$ when $t_L < 0$. Fig. 5 shows the evolution of the system starting from two different initial states for the three cases.

In another set of simulations, shown in detail in the attached technical report, the master input $\pi_{A1}(t)$ is chosen as in (19) to bring the load in $\tilde{p}'_L = [4.5, 4.5, 5]$ [m]. We observed that, as expected, for both $t_L = t_{L,1}$ and $t_L = t_{L,2}$ the system remains stable during the master maneuver. Once the input becomes constant, the master stops and the system converges to $\tilde{q}$ for $t_L = t_{L,1}$. For $t_L = t_{L,2}$, the final load attitude depends on the particular motion, and it is in general different from $\tilde{q}$.

Additional simulations in non-ideal conditions are provided in the attached technical report. The results show that thanks to the passivity of the system, the latter is very robust to the considered non-idealities. Some representative simulations are available in the attached video too.

**VII. CONCLUSIONS**

This work deals with the decentralized cooperative manipulation of a cable-suspended load performed by two aerial vehicles. The proposed master-slave architecture exploits an admittance controller in order to coordinate the robots with implicit communication only, exploiting the cable forces. The passivity of the system has been proven, and the stability of the static equilibria has been studied highlighting the crucial role of the internal force. In particular, contrarily from what

<table>
<thead>
<tr>
<th>System Parameters</th>
<th>Controller Gains</th>
</tr>
</thead>
<tbody>
<tr>
<td>$m_{B_0}$ [Kg]</td>
<td>$M_{B_0}$ 3F$_3$</td>
</tr>
<tr>
<td>$J_{0}$ [Kg · m$^2$]</td>
<td>0.015F$_3$</td>
</tr>
<tr>
<td>$l_{0}$ [m]</td>
<td>1</td>
</tr>
<tr>
<td>$h_{0}$ [N/m]</td>
<td>20</td>
</tr>
<tr>
<td>$b_{0}$ [m]</td>
<td>[0.433 0.0]</td>
</tr>
<tr>
<td>$m_{L,1}$ = 0.900 [Kg]</td>
<td>$J_{L,1}$ = 0.112 [Kg · m$^2$]</td>
</tr>
<tr>
<td>$\phi = 0$, $\theta = \pi/8$ [rad]</td>
<td>$\psi = \pi/7$ [rad]</td>
</tr>
</tbody>
</table>

**TABLE I:** Parameters used in the simulations.
it is normally done in the literature (zero internal force), it is advisable to choose a positive internal force to control both position and orientation of the beam. In the future it would be interesting to test the method on real platforms and to extend the analysis to general loads or to agile motions. An extension to a more generic load attached to \( N \) robots could be very interesting too.

**REFERENCES**


Additional Analysis and Simulations for Communication-Less Cooperative Aerial Manipulation

Technical report of:
“Aerial Co-Manipulation with Cables: The Role of Internal Force for Equilibria, Stability, and Passivity”
IEEE Robotics and Automation Letters, Special Issue on Aerial Manipulation

Marco Tognon¹, Chiara Gabellieri¹¹, Lucia Pallottino², and Antonio Franchi³

Abstract—This document is a technical attachment to [1] as an extension of the theoretical analysis and of the numerical validation part. Here we present additional plots and additional simulations in presence of non-ideal conditions as noise and parameter variations. A thorough validation of the robustness of the proposed method against the aforementioned non-idealities is also conducted.

I. HOW TO CITE THIS WORK

This technical report is accompanying our IEEE Robotics and Automation Letters paper [1]. If you wish to reference this work, please cite this paper as follows:

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  author = {M. Tognon and C. Gabellieri and L. Pallottino and A. Franchi},
  title = {Aerial Co-Manipulation with Cables: The Role of Internal Force for Equilibria, Stability, and Passivity},
  journal = {{IEEE} Robot. Autom. Lett.},
  Special Issue on Aerial Manipulation},
  year = {2018},
  doi = {10.1109/LRA.2018.2803811}
}

II. INTRODUCTION

In Sec. III we integrate the Lyapunov based stability analysis of Sec. IV of the manuscript. In particular we shall prove that the Lyapunov function (18) of the manuscript is positive definite and has global minima in a particular set of interest.

Then, in the following sections, we shall describe several additional simulations validating the proposed method and all the theoretical concepts and results presented in [1]. The parameter of the simulated system are the one reported in [1].

III. LYAPUNOV FUNCTION CHARACTERIZATION

In the following proposition we analyze in details the Lyapunov candidate:

\[ V(x) = \frac{1}{2}(v_R^T M_A v_R + e_R^T K_e e_R + v_L^T M_L v_L + k_1(||l_1|| - l_0) + k_2(||l_2|| - l_0^2) - l_{0}^2 f_1 + l_{0} \beta_2 f_2 + l_1 (1 - (\bar{R}_L e_1)^T R_L e_1) + V_0, \]

used in Sec. IV and Sec. V of the manuscript, for the proofs of stability and passivity of the system.

Proposition 1. Considering the Lyapunov function (1), we have that:

- \( x_{\text{min}} = \arg \min_{x} V(x) \) is such that \( x_{\text{min}} \in \mathcal{E}(0, \bar{q}_L) \) and \( x_{\text{min}} \in \mathcal{E}^+(l_i, \bar{q}_L) \) for \( l_i > 0 \);
- \( V(x) \) is positive definite for an opportune choice of \( V_0 \).

Proof. We divide (1) into three parts such that

\[ V(x) = \tilde{V}(x) + V_1(x) + V_2(x), \]

where

\[ \tilde{V}(x) = \frac{1}{2}(v_R^T M_A v_R + e_R^T K_e e_R + v_L^T M_L v_L + l_1 (1 - (\bar{R}_L e_1)^T R_L e_1) + V_0, \]

\[ V_i(x) = \frac{1}{2}k_i(||l_i|| - l_0) - l_{0}^2 f_1, \]

for \( i = 1, 2 \).

We firstly show that the Lyapunov function is radially unbounded (also called coercive), i.e., \( \lim_{||x|| \to \infty} \tilde{V}(x) = \infty \).

Indeed, we have that clearly \( \lim_{||x|| \to \infty} V_i(x) = \infty \), while

\[ \lim_{||x|| \to \infty} V_i(x) = \frac{1}{2}k_i(||l_i|| - l_0) - l_{0}^2 f_1 = 0 \]

\[ \lim_{||x|| \to \infty} V_i(x) = \frac{1}{2}k_i(||l_i||^2 + l^2_{0i}) - k_i||l_i||l_0 - l_{0}^2 f_1 \geq 0 \]

\[ \lim_{||x|| \to \infty} V_i(x) = \frac{1}{2}k_i(||l_i||^2 + l^2_{0i}) - k_i||l_i||l_0 - l_{0}^2 f_1 \geq 0 \]

\[ \lim_{||x|| \to \infty} ||l_i||^2 (k_i \frac{1}{2} + \frac{k_i^2 l_0^2}{2||l_i||^2} - \frac{k_i l_0}{||l_i||} - \frac{||f_i||}{||l_i||}) = +\infty. \]

Based on this results and on Theorem 1.15 of [2], we can say that function (3) has a global minimum. Now we can look for this minimum among the stationary points, i.e., where the
gradient $\nabla V(x) = 0$, and among the points where (1) is not differentiable [3].

It is clear that $\nabla \tilde{V}(x) = 0$ only if $v = 0$, $p_{R1} = \tilde{p}_{R1}$ and $t_L R_L e_1 \times \tilde{R}_L e_1 = 0$. Regarding $V_i(x)$, let us consider its gradient with respect to the cable configuration $l_i$:

$$\nabla_i V_i(x) = \frac{\partial V_i(x)}{\partial l_i} = k_i(||l_i|| - l_{0i}) \frac{l_i^T}{||l_i||} - \bar{f}_i. \quad (5)$$

Then, $\nabla_i V_i(x) = 0$ if and only if

$$k_i(||l_i|| - l_{0i}) \frac{l_i^T}{||l_i||} = f_i = \bar{f}_i. \quad (6)$$

Condition (6) holds in two different cases:

a) $k_i(||l_i|| - l_{0i}) = ||\bar{f}_i||$ and $l_i = \frac{\bar{f}_i}{||\bar{f}_i||}$, for which $||l_i|| > l_{0i}$ and $l_i^T \bar{f}_i > 0$;

b) $k_i(||l_i|| - l_{0i}) = -||\bar{f}_i||$ and $l_i = -\frac{\bar{f}_i}{||\bar{f}_i||}$, for which $||l_i|| < l_{0i}$ and $l_i^T \bar{f}_i < 0$.

The previous two cases have a straightforward physical interpretation. Since the cables are modeled as a spring, they can produce a force at a certain point both being stretched in the same direction of the force itself, as in case a), or being compressed in the opposite direction, as in case b). However in this work we consider only case a) because case b) is not practically feasible for cables, thus out of our region of interest.

Therefore $\nabla V(x) = 0$ if $x \in \mathcal{X}_{\tilde{V}_0} \cup \mathcal{X}_{V_0}$ where

a) $\mathcal{X}_{\tilde{V}_0} = \{x | v = 0, p_{R1} = \tilde{p}_{R1}, t_L R_L e_1 \times \tilde{R}_L e_1 = 0, k_i(||l_i|| - l_{0i}) = ||\bar{f}_i||, l_i = \frac{\bar{f}_i}{||\bar{f}_i||}\}$

b) $\mathcal{X}_{V_0} = \{x | v = 0, p_{R1} = \tilde{p}_{R1}, t_L R_L e_1 \times \tilde{R}_L e_1 = 0, k_i(||l_i|| - l_{0i}) = -||\bar{f}_i||, l_i = -\frac{\bar{f}_i}{||\bar{f}_i||}\}$

However, $x \in \mathcal{X}_{V_0}$ can not be the global minima since we can show that $V(x_a) < V(x_b)$ with $x_a \in \mathcal{X}_{\tilde{V}_0}$ and $x_b \in \mathcal{X}_{V_0}$. This comes from the fact that in (1), $-l_i^T \bar{f}_i < 0$ and $-l_i^T \bar{f}_i > 0$ for $x \in \mathcal{X}_{\tilde{V}_0}$ and $x \in \mathcal{X}_{V_0}$, respectively.

Finally we have to check the non-differentiable points of (1), namely the state $x \in \mathcal{X}_{l_{0i}} = \{x | ||l_i|| = 0 \text{ for } i = 1, 2\}$. Notice that this condition is out of our domain of interest. Nevertheless, also in this case we can show that $V(x_a) < V(x_{l_{0i}})$. Indeed, $V(x_{l_{0i}}) = V(x_{l_{0i}})$ and

$$V_i(x_{l_{0i}}) = \frac{1}{2} k_i l_{0i}^2$$

Thus $V_i(x_{l_{0i}}) < V_i(x_{l_{0i}})$.

For the last point, let us define the function $V'(x)$ as in (1) but without $V_0$. Then, in order to obtain $V'(x) > 0$, we can simply set

$$V_0 > \min_x V'(x) = V'(x_a)$$

with $x_a \in \mathcal{X}_{\tilde{V}_0}$.
IV. LOAD POSE REGULATION

Given the desired load configuration of equilibrium \( \vec{p}_L = [0.3 \ 0.3 \ 0.2]^T \text{[m]}, \vec{w} = \pi/7[\text{rad}] \) and \( \vec{\theta} = \pi/8[\text{rad}] \), we performed several simulations with \( \vec{p}_L \in \Pi_A(q_L) \) computed for the cases: 1) \( t_L = 1.5 > 0 \), 2) \( t_L = 0 \), 3) \( t_L = -1 < 0 \).

To study the stability of the equilibrium configuration for the different values of \( t_L \), we initialized the system in two different initial configurations and we observed the evolution of the system. We observed that for case

1) the system final configuration belongs to \( \mathcal{X}^+(t_L, q_L) \). Figure 1 shows the system configuration evolution for the two different initial conditions. The final state of the two trajectories is the same;

2) the system final configuration belongs to \( \mathcal{X}(0, q_L) \), and depends on the system initial state. Figure 2 shows the system configuration evolution in these cases;

3) the system does not converge to \( \mathcal{X}^-(t_L, q_L) \) even initializing it very close. Figure 3 shows the system configuration evolution in these cases.

The previous plots integrate the ones provided [1] showing the complete evolution of the main quantities of the system for two particular initial conditions.

![Fig. 1: Evolution of the system variables for \( t_L = 1.5 \text{ [N]} \) starting from two different initial conditions. The positions of the robots center of masses (which coincide with the cables attaching points) are shown, together with the position of the load center of mass and its yaw and pitch angles. The reference signals are displayed with dotted lines of the same color.](image-url)
Fig. 2: Evolution of the system variables for $t_L = 0$ [N] starting from two different initial conditions.

Fig. 3: Evolution of the system variables for $t_L = -1$ [N] starting from two different initial conditions.
V. LOAD TRANSPORTATION

Considering a time-varying control input, we defined \( \pi_{A1}(t) \) such that the master robot follows a 5th-order polynomial trajectory in the three directions (rest to rest with condition of zero acceleration at the initial and final points) starting from an initial position of \([1.18 \ 0.72 \ 2.2]' \text{[m]}\). The trajectory covers 4[m] along each of the three directions in 30[s]. The particular \( \pi_A(t) \), with \( \pi_{A2}(t) = \bar{\pi}'_{A2} \) and \( \pi_{A1}(t_f) = \bar{\pi}'_{A1} \), brings the load in the configuration \( \bar{\rho}_L = [4.5 \ 4.5 \ 5.0]' \text{[m]} \), \( \bar{\psi} = \pi/9 \text{[rad]} \) and \( \bar{\theta} = \pi/8 \text{[rad]} \). In Fig. 5 we show the results of the simulations in ideal conditions. We notice that, once the final input \( \pi_{A1}(t_f) = \bar{\pi}'_{A1} \) with \( t_L > 0 \text{[N]} \) is set, the system successfully transports the load between the two points stopping at the desired configuration, as shown in Fig. 4(a). For \( t_L = 0 \text{[N]} \) instead, the final load attitude depends on the particular motion, and it is in general different from the desired one, as shown in Fig. 4(b). Finally, as one can see in Fig. 4(c), when \( t_L < 0 \text{[N]} \) the final configuration of the system does not correspond to the desired one, since it was an unstable equilibrium. Notice in Fig. 4(a) how the error on the load trajectory remains sufficiently small for all the transportation, and goes to zero at the end of the task. In Fig. 5(a), 5(b) and 5(c) we show the results for a similar task, for \( t_L > 0 \), \( t_L = 0 \) and \( t_L < 0 \), respectively. In this case the trajectory is followed at a higher speed, since it is completed in 4 [s]. Consequently, as one can see in Fig. 5(a), the system moves faster and the tracking error increases. However it remains always bounded and the stability during the transportation is preserved. Furthermore, one could tune the admittance controller parameters of the slave robot to achieve better results if needed.

![Fig. 4: Evolution of the system variables during transportation for the three different values of internal force.](http://ieeexplore.ieee.org/)

Fig. 5: Evolution of the system variables during transportation for the three different values of internal force.
VI. NON-IDEAL CONDITIONS

In the following, we test the robustness of the proposed method against noise in the measured state and model parameter uncertainties. The following simulations consider the transportation scenario presented in Sec. V, where the trajectory is performed in 4[s].

A. Noisy Measurements

In Fig. 6 we report the results of a simulation where Gaussian noise is added to the estimated state of the robots and to the measured cable force, in order to simulate real sensors. In particular, the noise variances on the aerial vehicle position, velocity and measured cable force are equal to 0.005[m], 0.01[m/s] and 0.01[N], respectively. From the plots one can see that, even in the presence of noise, the system is able to bring the load to the desired pose showing only very small oscillations.

B. Noisy Measurements and Parametric Uncertainties

In Fig. 7 both measurement noise and parametric uncertainties are considered. In particular, the rest length of the cables, the cables anchoring points positions with respect to the center of mass of the load (or equivalently the position of the center of mass of the load) and the mass of the load are uncertain parameters. In other words, we put ourselves in a condition in which the real parameters and the nominal ones do not perfectly match. In particular, the known cables rest length has been set 5% greater than the real one, the load mass used to generate the constant control input $\pi A$ is 20% greater than the real one, and the anchoring points positions in body frame have been chosen as follows: $b_1 = [0.5 0.5 1]^{\top}$[m], $b_2 = [-0.47 0.02 0.03]^{\top}$[m]. With this simulation we want to show that the proposed algorithm is robust to uncertainties on the parameters in the sense that the system final equilibrium will be clearly different from the desired one, but the robots are still perfectly capable of performing the object transportation task in a stable way, as guaranteed by the system passivity. Fig. 7 shows the results of the simulation during the transportation. As one can see, the passive nature of the closed loop system makes the system state and output completely stable and converging to a constant equilibrium, that is of course different from the desired one because of the wrong parameter used. An adaptive approach could be used to reduce the effect of this.

C. Sensitivity to Load Mass Uncertainty

We performed several simulations varying the mass of the load known by the controller with respect to the real mass. Figure 8 displays how the load position and attitude errors at steady state, $e^p_L$ and $e^\theta_L$, change when the real mass is not exactly known. In particular, the errors are defined as:

$$e^p_L = ||p_L - \bar{p}_L||$$
$$e^\theta_L = ||\theta - \bar{\theta}|| + ||\psi - \bar{\psi}||.$$

The load starts from the configuration given by: $p_L(0) = [0.5 0.5 1]^{\top}$[m], $\psi = \pi/10$[rad], $\theta = \pi/8$[rad] and $m_L = 1.5$[N]. The desired final configuration is given by $\bar{p}_L = [0.5 0.5 1]^{\top}$[m], $\bar{\theta} = \pi/9$[rad], $\bar{\psi} = \pi/8$[rad], $m_L = 1.5$[N]. Calling $m'_L$ the known mass, we compute it as $m'_L = \Delta m \cdot m_L$, where $\Delta m$ is the relative mass increment. Figure 8 shows $e^p_L$ and $e^\theta_L$ with respect to $\Delta m$. The larger the parametric uncertainty on the load mass, the more the errors increase, too. However one can notice that even with an uncertainty grater than the 25% the system still remains stable. After the value $\Delta m = 1.3$ the system becomes unstable. Nevertheless, we remark that the mass of the load is one of the parameters that can be known with very good precision, also using an online estimation algorithm.

D. Sensitivity to Anchoring Point Position Uncertainty

As an additional study of the robustness of the proposed method, in Fig. 9 we show the load position and attitude errors at steady state when the parametric uncertainty is on
Fig. 8: Load position and attitude errors when the load mass known by the controller differs from the real one.

the position of the cables anchoring points on the load. In particular, the known anchoring positions are given by

\[
L_1 b_0 = L_2 b_1 + \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \Delta b \|L_1 b_1\|
\]

\[
L_2 b_2 = L_2 b_2 + \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix} \Delta b \|L_1 b_1\|,
\]

where \(\Delta b \in \mathbb{R}_{>0}\). The system starts from the configuration given by \(p_L(0) = [0.5 \ 0.5 \ 1]^T \text{[m]}, \ \psi = \pi/10 \text{[rad]}, \ \theta = \pi/8 \text{[rad]} \) and \(t_L = 1.5 \text{[N]}\). The desired final configuration is given by \(\bar{p}_L = [0.5 \ 0.5 \ 1]^T \text{[m]}, \ \bar{\theta} = \pi/9 \text{[rad]}, \ \bar{\psi} = \pi/8 \text{[rad]}, \ \bar{t}_L = 1.5 \text{[N]}\). Also in this case, as expected, the larger the parametric uncertainty on the considered quantities, the more the errors increase, too. However, with the considerable variation of \(\Delta b = 0.5\) the system still remains stable.

Fig. 9: Load position and attitude errors when the anchoring points position on the load known by the controller differs from the real one.
thrust required when the internal force increases. Given a

VII. EFFECTS OF THE INTERNAL FORCE ALONG THE LOAD

In the manuscript we saw that to make a desired load configuration asymptotically stable, one has to compute the proper control input \( \pi_A(\dot{q}, t_L) \) choosing \( t_L > 0 \). In the following, we shall analyze the effects of the intensity of the internal force on the system behavior. In this way we can better decide the value of \( t_L \). In particular, in the following, we shall analyze the relations between \( t_L \) and convergence time, and between \( t_L \) and required total thrust in the equilibrium configuration.

A. Internal Force and Convergence Time

If the internal force \( t_L = 0 \) [N] the load does not in general converge to its desired pose, which instead happens for \( t_L > 0 \). However, it is interesting to see how the convergence rate behaves changing the intensity of the internal force. In Fig. 10, we show how the convergence time of the load position and attitude, defined by \( t_c \), varies when increasing the internal force. Here \( t_c = \min\{t_c^e, t_c^p\} \), where \( t_c^e \) is the time after which \( \epsilon^e_L \) remains below 5°, while \( t_c^p \) is the time after which \( \epsilon^p_L \) remains below 0.02 [m]. The initial and the final desired load configurations are the same as before. Notice that for \( t_L = 0 \) the convergence time is in general infinite. One can notice that increasing \( t_L \) up to 0.7 [N], \( t_c \) decreases. However, after this value, \( t_c \) starts to increase due to the appearance of some larger oscillations that takes more time to be damped. In any case, this study shows that even a minimal internal force of 0.1 [N] is enough to obtain asymptotically stability for which an almost negligible increase of total thrust is required.

B. Internal Force and Total Thrust

Since the internal force, necessary to make the load converge to the desired pose, implies an added energy consumption for the robots, we evaluated the amount of additional thrust required when the internal force increases. Given a certain desired load pose, Fig. 11 shows the relative increase of total thrust, \( \Delta f_R \), augmenting the intensity of the internal force with respect to the total thrust required by the case with zero internal force. In particular \( \Delta f_R \) is computed as

\[
\Delta f_R(t_L) = f_R(t_L) - f_R^0,
\]

where \( f_R(t_L) \) is the the sum of the thrusts required by the two vehicles at steady state to stabilize the load at a certain load configuration with a certain value of \( t_L \), and \( f_R^0 = f_R(0) \).

One can notice that even imposing \( t_L = 1 \) [N], much higher than the real internal force required to stabilize the system, the \( \Delta f_R \) is below the 0.005, i.e., the total extra thrust is lower than the 0.5% of the total thrust required with \( t_L = 0 \) [N].

In any case we remark that the proposed control method is still applicable for \( t_L = 0 \) [N]. The system is proven to be still stable, but will not clearly asymptotically converge to the desired pose.

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