Immersion and Invariance Control for an Antagonistic Joint with Nonlinear Mechanical Stiffness

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Abstract—Tendon controlled robots with nonlinear mechanical tendon stiffness are becoming more and more popular. With the appropriate actuation, the joint position and the stiffness between motor and link side can be prescribed at the same time. In this paper, the modeling of tendon-driven elastic systems with nonlinear couplings is recapitulated. Based on the Immersion and Invariance (I&I) framework a control law is developed that takes trajectories of the desired joint position and the pretension as input. The contribution of this paper are the application of the I&I framework to tendon-controlled systems with variable stiffness which requires to consider the internal degrees of freedom to realize the pretension, respectively the mechanical joint stiffness. But also, and even more importantly, the handling of nonlinear transmission elements between the motor and the joint link is included. The performance and sensitivity is analyzed by simulating an antagonistically driven robot joint.

I. INTRODUCTION

The antagonistic tendon actuation principle from biology can be applied to robots by using tendons that connect the joints with the motors allowing a relocation of the drives to the base of the robot system. This design furthermore reduces the inertia of the robot system leading to faster motion while preserving safety for the human and the robot. A limitation in such systems is that excessively large tendon forces might occur in case of stiff tendons, which in the worst case leads to their breaking. The introduction of elastic elements into the tendon path relaxes this problem. This was done in some developments using linear tendon stiffness and by the introduction of variable stiffness (VS) elements [1]–[6]. Another example is the DLR hand arm system depicted in Figure 1 [7]. One of the first commercially available light-weight arms, namely the WAM [8], [9], is also based on tendon actuation. Controllers were presented that are capable of keeping the tendon tension while moving the robot joints with linear tendon stiffness [10]–[13]. A serial stiffness acts furthermore as mechanical low-pass filter to external forces, e.g. in case of a collision. In contrast to a constant elastic element, a variable stiffness allows to adapt the joint compliance to a given manipulation task, like for catching a heavy object in order to prevent damage of the robot or for fast assembly of parts. Another feature is that the serially connected springs act as energy storage that can be applied to e.g. a throwing task through transferring the stored spring energy to kinetic energy and vice versa [6], [14]. Furthermore, in [15] an impedance control law for a VS joint was developed. A tracking controller has been proposed and has been validated using a linearly coupled mechanism through experiments [16]. In [5] Palli et al. studied the feedback linearization of uncoupled joints that are each driven antagonistically. Furthermore, an adaptive motor level PD controller was presented in [17] for a single antagonistic joint that was verified by means of experiments. Passivity-based impedance controllers for flexible joint robots were presented by Ott and Albu-Schäffer [18], [19]. In this framework, the emphasis was to derive controllers from potential functions and to develop control laws that realize a predefined stiffness characteristics in the joints by means of control. In [20] we presented a particular PD controller that can set joint positions and joint stiffness independently in a robust fashion. However, since the controller was only based on the motor positions, the transient behavior could not fully be specified. The contribution of this paper are the application of the I&I framework [21] to tendon-controlled systems with variable stiffness which requires the handling of nonlinear transmission elements between the motor and the joint link, and to consider the internal degrees of freedom that are related to pretension, respectively effective mechanical joint stiffness.

In the second section we will first review the I&I stabilization theorem as proposed by Astolfi et al. in [21]. The modeling of an antagonistic joint with nonlinear tendon stiffness is described in section 3, while the I&I control is applied in section 4. The developed controller is compared in simulation with a motor PD controller in section 5.

II. I&I STABILIZATION

This section introduces the I&I framework with the main theorem proposed by Astolfi et al. in [21].
Theorem 1: Consider the system
\[ \dot{x} = f(x) + g(x)u, \]
with \( x \in \mathbb{R}^n \), \( u \in \mathbb{R}^m \), and an equilibrium point \( x^* \in \mathbb{R}^n \) to be stabilised. Assume that there exist smooth mappings \( \alpha : \mathbb{R}^p \to \mathbb{R}^p, \pi : \mathbb{R}^p \to \mathbb{R}^n, \phi : \mathbb{R}^n \to \mathbb{R}^{n-p}, c : \mathbb{R}^p \to \mathbb{R}^m \) and \( v : \mathbb{R}^{n \times (n-p)} \to \mathbb{R}^m \), with \( p < n \), such that the following hold.

1) The target system
\[ \dot{\xi} = \alpha(\xi), \]
with \( \xi \in \mathbb{R}^p \) has a globally asymptotically stable equilibrium \( \xi^* \in \mathbb{R}^p \) and
\[ x^* = \pi(\xi^*). \]

2) For all \( \xi \in \mathbb{R}^p \),
\[ f(\pi(\xi)) + g(\pi(\xi))c(\pi(\xi)) = \frac{\partial \pi}{\partial \xi} \alpha(\xi). \]

3) The set identity
\[ \{x \in \mathbb{R}^n | \phi(x) = 0\} = \{x \in \mathbb{R}^n | x = \pi(\xi), \xi \in \mathbb{R}^p\} \]
holds.

4) All trajectories of the system
\[ \dot{z} = \frac{\partial \phi}{\partial x}(f(x) + g(x)v(x, z)) \]
\[ \dot{x} = f(x) + g(x)v(x, z), \]
are bounded and (5) has a uniformly globally asymptotically stable equilibrium at \( z = 0 \).

Then \( x^* \) is a globally asymptotically stable equilibrium of the closed-loop system
\[ \dot{x} = f(x) + g(x)v(x, \phi(x)). \]

\[ \square \]

Taken from [21], this theorem can be interpreted in the following way. The objective is to find a manifold \( \mathcal{M} = \{x \in \mathbb{R}^n | x = \pi(\xi), \xi \in \mathbb{R}^p\} \) based on the system (1) and the target dynamics (2). This manifold can be rendered invariant and asymptotically stable, and such that the well-defined restriction of the closed-loop system to \( \mathcal{M} \) is described by the target system (condition 3). Note that the control input \( u \) that makes the manifold invariant is not unique, since it is uniquely defined only on \( \mathcal{M} \). One possible control, that drives the off-the-manifold coordinates \( z \) to zero and keeps the system bounded, is selected. The I&I concept is illustrated for \( p = 2 \) and \( n = 3 \) in Fig. 2.

### III. Modeling of an Antagonistic Joint with Nonlinear Tendon Stiffness

In Fig. 3, a simple tendon-network consisting of one joint and two tendons connected by nonlinear springs is shown. In Table I, the variables to describe the equations of motion of an antagonistic robot joint are given. The tendon inverse kinematics \( h_q(q) \) gives the motor positions as a function of the joint angles \( q \). The function \( h_q(q) \) can be used to derive a differential map \( P(q) \):
\[ P(q) = \left( \frac{\partial h_q(q)}{\partial q} \right)^T. \]

In the robotics literature, this map is also known as **coupling matrix** [22]. Note that in contrast to the Jacobian matrix of a serial kinematic chain, the transposed coupling matrix maps from joint to tendon velocities
\[ \dot{h}_q = P^T(q)\dot{q} \]
and the tendon forces are related to the joint torques by
\[ \tau_q = P(q)f_t. \]

For a well-designed mechanism, the coupling matrix \( P(q) \) has full row rank over the whole configuration space [23]. Using the coupling matrix \( P(q) \), it is straightforward to formulate the equations of motion for an antagonistic joint.
with \( K_p, K_d > 0 \), \( \xi^*_i(t) \in C^4 \).

### B. Computation of the map \( \pi(\xi, t) \)

A major difficulty using the I&I framework is to solve the partial differential equation (3). Since \( \ddot{w}(\xi, t) \) is an explicit function of time, additionally the term \( \frac{\partial \pi}{\partial t} \) has to be included

\[
f(\pi(\xi, t)) + g(\pi(\xi, t))c(\pi(\xi, t)) = \frac{\partial \pi}{\partial \xi} \alpha(\xi, t) + \frac{\partial \pi}{\partial t} \xi.
\]  

Since the target system (16) resembles the link side dynamics of the original system (15), the solution to this equation can be obtained by starting to set \( \pi_1(\xi) = \xi_1 \). Then, looking at \( \dot{x}_1 = x_2 \) it follows directly that \( \pi_2(\xi) = \xi_2 \). Based on the term \( \dot{x}_2 \), the mapping \( \pi_3(\xi, t) \) is derived as

\[
\dot{x}_2 = \frac{\partial \pi_2(\xi)}{\partial \xi} \dot{\xi},
\]

using the systems (15) and (16)

\[
M^{-1}(P f_1(\xi_1, \pi_3(\xi, t)) + \tau_{ext}) = -M^{-1}(\ddot{w}(\xi, t) - \tau_{ext}),
\]  

with \( P \) tendon-controllable [23] and therefore with full row rank. This equation is rewritten as

\[
f_1(\xi_1, \pi_3(\xi, t)) = -P + \ddot{w}(\xi, t) - N(P) w_n(t) = -w(\xi, t),
\]  

with the nullspace basis of \( P \), \( N(P) = -\sqrt{2} [1, 1]^T, \)

\( w_n(t) \in C^2, \)

and \( w(\xi, t) \) the modified rigid body controller. This furthermore gives the possibility to use the nullspace of \( P \) to add feed-forward forces \( w_n \) that shall realize pretension between the motors while at the same time not affecting the link motion. Note, that we assume in this paper that the pretension \( w_n \) is set such that the pulling constraint for the tendons is fulfilled. In the case of nonlinear mechanical stiffness this will be used to specify a desired mechanical stiffness between the motors and the robot link [1], [20].

This set of nonlinear equations needs to be solved for \( \pi_3(\xi, t) \). Since in each equation only the \( i^{th} \) component of \( \pi_3(\xi, t) \) is a function of the known terms, the solution will be derived componentwise. Using the force model \( (14) \), the \( i^{th} \) component of \( \pi_3(\xi, t) \) can be computed to

\[
\pi_{3,i}(\xi, t) = (P^T\xi)_i + \gamma_i^{-1} \ln(1 - k_{i,i}^{-1} w_i(\xi, t)).
\]  

The solution to \( \pi_3(\xi) \) can then be given by stacking the components

\[
\pi_3(\xi, t) = P^T \xi_1 + \begin{pmatrix} \gamma_1^{-1} \ln(1 - k_{1,1}^{-1} w_1(\xi, t)) \\ \gamma_2^{-1} \ln(1 - k_{1,2}^{-1} w_2(\xi, t)) \end{pmatrix}.
\]

The expression \( \pi_3(\xi) \) can be interpreted as virtual desired motor positions that are required to prescribe the target dynamics.

Using \( \dot{x}_3 = x_4 \) appearing in the system equations (15) it is straightforward to derive the map \( \pi_4(\xi, t) \), respectively its \( i^{th} \) component as

\[
\pi_{4,i}(\xi, t) = \frac{\partial \pi_{3,i}(\xi, t)}{\partial \xi} \dot{\xi} + \frac{\partial \pi_{3} (\xi, t)}{\partial t},
\]  

\[i\text{Certainly, the tracking controller proposed by Slotine and Li [24] could be applied as well.}
while keeping in mind that \( w(\xi, t) \) is also a function of time, and with

\[
\frac{\partial \pi_3, i(\xi, t)}{\partial w_i} = \frac{-\gamma_i k_i^{-1}}{1 - k_i^{-1} w_i(\xi, t)}. \tag{24}
\]

Note that the denominator becomes zero at \( w_i(\xi, t) = k_i \). This corresponds to the force \( f_{i, i}(x_1, x_2, i) = -k_i \), which means that \( x_3, i - (P x_1, i) \rightarrow -\infty \), and is not relevant here since (positive) pulling forces are assumed. Together with the system equations (15) the solution can be computed as

\[
\pi_4, i(\xi, t) = (P^T \xi_2) + \frac{\partial \pi_3, i(\xi, t)}{\partial w_i} \left( \frac{\partial w_i(\xi, t)}{\partial \xi_1} \xi_2 \right) \left( M^{-1}(P w(\xi, t) - \tau_{ext}) \right) - \frac{\partial w_i(\xi, t)}{\partial \xi_2} \left( M^{-1}(P w(\xi, t) - \tau_{ext}) \right) + \frac{\partial w_i(\xi, t)}{\partial \tau_{ext}} \left( M^{-1}(P w(\xi, t) - \tau_{ext}) \right). \tag{25}
\]

C. Derivation of the manifold \( \phi(x, t) = 0 \)

The next step is to find an explicit definition of the manifold \( \phi(x, t) \in \mathbb{R}^{6-2} \), i.e.,

\[
\phi(x, t) = \begin{pmatrix} \phi_{1, 1}(x, t) \\ \phi_{1, 2}(x, t) \\ \phi_{2, 1}(x, t) \\ \phi_{2, 2}(x, t) \end{pmatrix}. \tag{26}
\]

With the direct correspondences of \( \pi_1(\xi) = x_1, \pi_2(\xi) = x_2, \) two equations of the manifold are constructed by

\[
\phi_1(x, t) = x_3 - \pi_3(\xi, t)|_{\xi_1 = x_1, \xi_2 = x_2} \tag{27}
\]

\[
\phi_1(x, t) = x_3 - P^T \xi_1 - \left( \frac{\gamma_i k_i^{-1}}{1 - k_i^{-1} w_i(\xi, t)} \right). \tag{28}
\]

In a similar fashion \( \phi_2(x, t) \) is constructed. First, \( \phi_2(x, t) \) is derived as

\[
\phi_2(x, t) = x_4 - \pi_4(\xi, t)|_{\xi_1 = x_1, \xi_2 = x_2}. \tag{29}
\]

Secondly, as pointed out in [21], the already obtained set equation \( \phi_1(x, t) = 0 \) is used. Instead of using this function directly we rewrite it as \( w(x, t) = -f_{i, i}(x_1, \pi_3(x, t)) \) that is inserted in \( \phi_2(x, t) \). In this way, we achieve that \( \phi_1(x, t) = \phi_2(x, t) \) which is helpful to simplify the controller design to render the manifold attractive. Note that even though the term \( \pi_4(\xi, t) = \pi_3(\xi, t) \) appears in \( \phi_2(x, t) \), this not equal to \( \phi_2(x, t) = \phi_1(x, t) = \frac{\partial \phi_{i, x}}{\partial x} \cdot x \), since \( f_{i, i}(x_1, \pi_3(x, t)) \) does not appear in the target dynamics. The term \( \phi_2(x, t) \) is obtained for index \( i \) as

\[
\phi_{2, i}(x, t) = x_{4, i} - \pi_3, i(\xi, t), \tag{30}
\]

\[
\pi_{4, i}(\xi, t) = (P^T x_2) + \frac{\partial \pi_3, i(\xi, t)}{\partial \xi_1} \left( \frac{\partial w_i(\xi, t)}{\partial \xi_1} x_2 \right) \left( M^{-1}(P f_{i}(x_1, \pi_3(x, t)) + \tau_{ext}) \right) + \frac{\partial w_i(\xi, t)}{\partial \xi_2} \left( M^{-1}(P f_{i}(x_1, \pi_3(x, t)) + \tau_{ext}) \right). \tag{31}
\]

Having defined the manifold \( \phi(x, t) \), the equations of the off-the-manifold dynamics can be determined according to the modified equation (5)

\[
\dot{z} = \frac{\partial \phi(x, t)}{\partial x} (f(x) + g(x)v(x, z, t)) + \frac{\partial \phi(x, t)}{\partial t}. \tag{32}
\]

The equation of \( \dot{z}_1 = z_2 \) is obtained by the construction of \( \phi(x, t) \). With equation (29) the remaining system equations for \( \dot{z}_2 \) are determined for each index \( i \) by differentiation as

\[
\dot{z}_{2, i} = m_i(x, t) + (M^{-1})_{i, i} u_i, \tag{33}
\]

\[
m_i(x, t) = -\left(M_\theta^{-1}ight)_{i, i} (f_{i, i}(x_1, \pi_3(x, t)) + f_{i, 0, i}) - (P^T \dot{x}_2) \left( \frac{\partial \pi_3, i(\xi, t)}{\partial \xi_1} \right) \left( \frac{\partial w_i(\xi, t)}{\partial \xi_1} x_2 \right) \left( M^{-1}(P f_{i}(x_1, \pi_3(x, t)) + \tau_{ext}) \right) + \frac{\partial w_i(\xi, t)}{\partial \xi_2} \left( M^{-1}(P f_{i}(x_1, \pi_3(x, t)) + \tau_{ext}) \right). \tag{34}
\]

The control law is then chosen as

\[
u = v(x, z, t) = -M_\theta(m(x, t) + K_{z, 1} z_1 + K_{z, 2} z_2), \tag{35}
\]

with \( K_{z, 1}, K_{z, 2} \in \mathbb{R}^{2 \times 2} \) which are each positive-definite matrices such that the off-the-manifold dynamics are stabilized. Note that the \( z_1 \) term represents the error term between the position of the motors and the desired virtual motor positions of the target system. This means the stabilization of the off-the-manifold dynamics can be interpreted as an underlying motor position controller.

D. Discussion of the boundedness of the closed-loop system

In this paragraph the boundedness of the closed-loop equations with the states \((x, z)\) is analyzed in order to show that condition 4 of the I&I stabilization theorem holds. The closed loop equations are obtained by inserting the control law (31) in the system equation (15). Then, the (fast) coordinates \( x_3 \) and \( x_4 \) are replaced by the coordinate transformation \( \phi_1 = x_3 - \pi_3(x, t) \) and \( \phi_2 = x_4 - \pi_4(x, t) \). To complete the change of coordinates we need to solve for \( x_3 = \phi_1 + \pi_3(x, t) \) that is inserted into the remaining systems equations

\[
\begin{align*}
\dot{z}_1 &= z_2 \tag{36} \\
\dot{z}_2 &= -K_{z, 1} z_1 - K_{z, 2} z_2 \tag{37} \\
\dot{x}_1 &= x_2 \tag{38} \\
\dot{x}_2 &= M^{-1}(P f_{i}(x_1, \phi_1 + \pi_3(x, t))) \tag{39} \\
\dot{\phi}_1 &= \pi_3 \tag{40} \\
\dot{\phi}_2 &= -K_{z, 1} z_1 - K_{z, 2} z_2, \tag{41}
\end{align*}
\]

while setting \( \tau_{ext} = 0 \) to zero for this analysis. By construction, the off-the-manifold dynamics converge exponentially. Then, from equation (34) and (35) it can be seen that \( \phi_1 \) is bounded. Therefore, it remains to show that the set of equations (34) and (35) has a bounded solution. We therefore consider these equations as system with nonvanishing perturbation and follow Lemma 9.2 in [25]. The perturbation is generated by the term \( \phi_1 \). Setting \( \phi_1 = 0 \), the set of equations that represents the linear target dynamics is obtained. The nominal system is then given as

\[
\begin{align*}
\dot{x}_1 &= x_2 \tag{42} \\
\dot{x}_2 &= -M^{-1} P w(x, t). \tag{43}
\end{align*}
\]
In order to bring the equations (34), (35) into a form to identify the total perturbation term $b_2(x_1, x_2, \phi_1, t)$ the term $M^{-1}Pw(x, t)$ from (39) is added and subtracted from equation (35) resulting in the perturbed system

\[
\begin{align*}
\dot{x}_1 &= x_2 \\
\dot{x}_2 &= -M^{-1}Pw(x, t) + b_2(x_1, x_2, \phi_1, t) + M^{-1}Pf_1(x_1, \phi_1 + \pi_3(x, t)).
\end{align*}
\]

Together with the tendon force model (14), the definition of $\pi_3(x, t)$ in (22) (that contains the inverse operation on the tendon force model), and the identity $e^{a+b} = e^a e^b$, the perturbation term can be reformulated as

\[
b_2(x_1, x_2, \phi_1, t) = -M^{-1}P \text{diag}\{\delta\}w(x, t) + M^{-1}PK_1\delta,
\]

with $\delta_i = e^{\gamma_i \phi_1} - 1$, for $i = 1, 2$, and the matrix $K_1 = \text{diag}\{k_{1,1}, k_{1,2}\}$. Note, that the specific structure of the perturbation term is due to the particular choice of force characteristics. Note that as $\phi_1 \to 0$ it follows that $b_2 \to 0$.

The linear nominal system (38), (39) is transformed to the coordinates

\[
e_1 = x_1 - \xi_1^*(t), \quad e_2 = x_2 - \xi_1^*(t).
\]

With these coordinates and inserting (17) the nominal system becomes

\[
\dot{e}_1 = e_2 \\
\dot{e}_2 = -M^{-1}(K_p e_1 + K_d e_2)
\]

For the nominal linear system there exists a Lyapunov function [25]

\[
V(e) = e^T P_L e,
\]

with $P_L A + A^T P_L = -Q = -I$ (43) and $A = \begin{bmatrix} 0 & 1 \\ -M^{-1}K_p & -M^{-1}K_d \end{bmatrix}$.

with $e = (e_1, e_2)^T$. This Lyapunov function will be used as a candidate potential function for the perturbed system (34), (35) for which we want to show boundedness. As intermediate step, the perturbation term $b_2$ (41) is rewritten in the coordinates $e$, i.e.

\[
b_2(e, \phi_1, t) = -M^{-1}P \text{diag}\{\delta\} P^+(K_p e_1 + K_d e_2) + M^{-1}P(K_1 \delta - \text{diag}\{\delta\} P^+ M \xi_1^*).
\]

The derivative of $V(e)$ along the trajectories of the perturbed system satisfies

\[
\dot{V}(e, t) = -\|e\|^2 + e^T P_L e + 2e^T P_L \delta,
\]

with $\delta = \begin{bmatrix} 0 & -M^{-1}P(K_1 \delta - \text{diag}\{\delta\} P^+ M \xi_1^*) \end{bmatrix}$. $K_\delta = P_L A_\delta + A_\delta^T P_L$, and

\[
A_\delta = \begin{bmatrix} 0 & 1 \\ -M^{-1}P \text{diag}\{\delta\} P^+ K_p & -M^{-1}P \text{diag}\{\delta\} P^+ K_d \end{bmatrix}
\]

For the quadratic perturbation term the inequality

\[
e^T K_\delta e \leq \lambda_{max}(K_\delta) \|e\|^2
\]

can be used. For the second (indfinite) term the Cauchy-Schwarz inequality $a^T b \leq \|a\| \|b\|$ is applied that gives

\[
2e^T P_L \delta \leq 2\|e^T P_L \|\|\delta\|\|\|e\|\|\|\delta\|\|.
\]

Using these two inequalities we obtain

\[
V(e, t) \leq -\|e\|^2 + \lambda_{max}(K_\delta) \|e\|^2 + 2\lambda_{max}(P_L) \|e\|\|\delta\|.
\]

Suppose $\lambda_{max}(K_\delta) \leq (1 - \zeta), \ z < 1$, we can write

\[
V(e, t) \leq -\|e\|^2 + (1 - \zeta)\|e\|^2 + 2\lambda_{max}(P_L) \|e\|\|\delta\|.
\]

Since $\delta$ is bounded and $P_L$ exists, we conclude that the solutions of the perturbed system are bounded. Hence, condition 4 of the I&I stabilization theorem is fulfilled while considering the restrictions on $\nu$ and $\zeta$. With the equations (22), (25), condition 2), and with the equations (27), (29) condition 3) of the theorem are fulfilled. Since condition 1) is valid by definition, $x^* = \pi(\xi^*)$ is a globally asymptotically stable equilibrium of the closed-loop system (15) with control law (31) that uses the modified rigid body control law (17), (20).

V. Simulations

For the simulations an antagonistic joint was considered. The plant parameters and the controller parameters are given in Table II with the friction model $f_{f, \theta} = f_{f, \theta} x_s$. The sample time was set to 0.1 ms. The I&I controller is compared with a simple motor PD controller with setpoints that are derived from the desired link position and desired pretension. Both controllers are tested to change the pretension $w_n(t)$ from 0 to $5\sqrt{2}$ Nm starting at time $t = 0$ s. At time $t = 2s$ a link side motion $\xi_1^*(t)$ to move sufficiently smooth from 0 rad to 0.2 rad is commanded. The reaction to a first order filtered step (time constant 10 ms) in external torque of $\tau_{ext} = 1$ Nm on the link was examined at time $t = 4s$. The trajectories of these input values are illustrated in Fig. 4. At the end of this section the sensitivity to parameter uncertainty is examined.

A. I&I Control

The positions of the link and the motors for this simulation for the I&I controller are depicted in Fig. 5. In the following presentation then only the tracking errors\footnote{This condition can only be fulfilled for sufficiently small values of $\delta_i < |\delta_{max}|$. This threshold grows by choosing smaller values for the controller parameters $K_p, K_d$.} will be discussed. At time $t = 0$, the pretension $w_n(t)$ is changed from 0 to $5\sqrt{2}$ Nm. It can be seen in the figures 5 and 6 that both the motors move in the same direction such that they wind up against each other, and in this way the pretension is realized. Since the tendon stiffness is nonlinear this pretension results in larger tracking error. The corresponding desired link position. Therefore, the motor displacements were computed as $P^T \xi_1^* - x_s$.\footnote{The displacements in the motor positions are represented w.r.t. the corresponding desired link position. Therefore, the motor displacements were computed as $P^T \xi_1^* - x_s$.}
as well as in an increased mechanical joint stiffness. In the case of the motion command at $t = 2s$, the tracking error, that can be observed in Fig. 6, remains below $3 \times 10^{-3}$rad which demonstrates the excellent tracking capabilities. This error stems from the gains of $K_z$. With increasing $K_z$, the tracking error decreases. Note that the acceleration feedforward term $M\ddot{\xi}_z(t)$ requires sufficiently smooth trajectories. Otherwise, it generates very large peaks for the control input, since its second derivative is used in equation (30). These peaks are generated to minimize the effect of the motor inertia and enables therefore the accurate tracking of the link position.

A filtered step in external torque of 1 Nm was exerted on the link at time $t = 4s$. A compliant behavior of the link is observed and behaves accordingly to the target dynamics system (16). For the steady-state the deflection is $0.002$rad that corresponds to the expected $\tau_{ext}/K_p = 1/500$rad. In Fig. 7 the control signal shows the additional component to realize the pretension. The off-the-manifold dynamics converge to zero (c.f. Fig. 8) and therefore the required virtual motor positions are realized. Note that for this parametrization the term $\delta$ stay by far below the threshold $|\delta_{max}| \approx 0.164$.

The performance of the controller is compared with a simple motor PD controller with setpoints that are derived from the desired link position and desired pretension

$$f_m = -K_{p,m}e_m - K_{d,m}e_m + M\ddot{x}_{3,d} + f_{f,d} + N(P)w_n \quad (52)$$

with $e_m = x_3 - P^T x_{3,d}$, and $K_{p,m} = \text{diag}(250, 250) Nm/rad, K_{d,m} = \text{diag}(25, 25) Nms/rad$ the motor stiffness and the motor damping matrices, respectively. The inertia matrix for the feedforward of the acceleration term $M_{m}$ is defined by the sum of the motor inertia $M_{m}$ and the reflected joint inertia $P^T M P^T$. The desired motor positions are computed by $x_{3,d} = P^T \xi_1 + \gamma_i^{-1} \ln(1 - k_{\xi_i}^{-1} w_n)$. In Fig. 9 it can be seen that the pretension is well adjusted. It can be further observed that the tracking error is by several magnitudes larger than the one of the I&I controller. The corresponding control signal $f_m$ is depicted in Fig. 10. Parallel to this controller we computed the off-the-manifold dynamics that are shown in Fig. 11 in which weakly damped oscillations

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**TABLE II**

<table>
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<tr>
<th>$M$</th>
<th>$1$</th>
<th>$kgm^2$</th>
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<tbody>
<tr>
<td>$M_\theta$</td>
<td>$0.5$</td>
<td>$0$</td>
</tr>
<tr>
<td>$P$</td>
<td>$[1 \quad -1]$</td>
<td>$m$</td>
</tr>
<tr>
<td>$\gamma_1$</td>
<td>$1$</td>
<td>$1/rad$</td>
</tr>
<tr>
<td>$k_{f,0}$</td>
<td>$500$</td>
<td>$Nms/rad$</td>
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<tr>
<td>$f_{f,0}$</td>
<td>$0.001$</td>
<td>$Nms/rad$</td>
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<tr>
<td>$K_p$</td>
<td>$500$</td>
<td>$Nm/rad$</td>
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<tr>
<td>$K_d$</td>
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<td>$Nms/rad$</td>
</tr>
<tr>
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<td>$25$</td>
</tr>
<tr>
<td>$K_{z,2}$</td>
<td>$25$</td>
<td>$100$</td>
</tr>
</tbody>
</table>

**Fig. 5.** I&I control: Link and motor positions. Note that in the pretension phase both motors move in the same direction while moving in the opposite direction in the case of moving the link position. The external torque displaces the link position as specified in the target dynamics. At the same time the motor positions are displaced as well.

**Fig. 6.** I&I control: Position error for link and motor positions.

**Fig. 4.** Trajectories of the commanded desired joint positions, of the pretension, and of the external torque.

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The motor controller gains have been chosen to fulfill the equations $K_p = PK_{p,m}P^T$ and $K_d = PK_{d,m}P^T$. 

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**Fig. 9.** Comparison of the I&I control with a simple PD motor control.
C. Sensitivity Analysis

The same task was performed again for the I&I controller, whereas the friction parameter, that is often not well known in practice, was replaced by an estimate $f_{f,0} = 1.3 f_{f,0}$, and at the same time the motor inertia was assumed to be $M_{\theta} = 0.7 M_{\theta}$ in order to study the sensitivity of the controller to parameter errors. Note, that the friction compensation term $f_{f,\theta}$ appears in the control law (31). In Fig. 12 it can be seen that the tracking error and the settling time is increased in the case of the link motion command at $t = 2.0s$. This indicates a robust behavior w.r.t. parameter uncertainties. In comparison the simulation results are shown for the case of the simple motor controller in Fig. 13. From that we defer that even in the case of (reasonable) parameter uncertainties the I&I controller has a better tracking performance.

VI. CONCLUSION

In this paper the Immersion and Invariance controller was applied to an antagonistic joint with nonlinear mechanical stiffness. The model of a tendon controlled variable stiffness mechanism was briefly introduced. The target system was specified to behave as desired second order impedance while at the same time a pretension was commanded to the tendon forces that lie in the nullspace of the tendon coupling matrix. In this way, we were able to assign on the one side the controlled impedance behavior for the link and at the same time set the mechanical stiffness of the mechanism by varying the pretension. Furthermore, we discussed the physical meanings of the map $\pi$ that corresponds to virtual desired motor positions and that the off-the-manifold stabilization represents an underlying motor position controller. The I&I controller was compared in simulations with a simple motor PD based controller. The I&I controller clearly outperforms the motor PD controller and the off-the-manifold coordinates are well damped and converge to zero - even in the case of parameter uncertainties. As future work we plan to implement this controller on hardware and to expand to the case of multiple degrees of freedom. Furthermore, we consider to use a reduced-order observer for the robotic system.

ACKNOWLEDGEMENT

This work has been partially funded by the European Commissions Seventh Framework Programme as part of
the projects VIACTORS (grant no. 231554) and The Hand Embodied (grant no. 248587).

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