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Robust control law strategy based on High Order Sliding Mode: towards a muscle control.

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Abstract—Functional electrical stimulation (FES) is used to excite paralysed muscles that would otherwise be uncontrollable by paraplegic patients. Consequently, the patient could recover partially some of lower limb functions improving the cardiovascular system, increasing oxygen uptake and bettering the whole quality of life. In this paper, we apply a control design based on a Higher Order Sliding Mode to a complex physio-mathematical muscle model. This model is based on macroscopic Hill and microscopic Huxley concepts. The main goal concerns the prediction of the needed pattern stimulation (current and pulse width), which will extend the overall performances and defer the muscle fatigue as much as possible. The controller is mathematically computed and shown to provide satisfactory stability and tracking errors. Its efficiency is illustrated with the control of the knee joint angle under a co-contraction approach.

Index Terms—FES, High order sliding mode, closed loop control, non linear system, muscle model.

I. INTRODUCTION

Functional electrical stimulation (FES) can help in regaining limited locomotor activities in humans with paraplegia through electrical stimulation of the lower extremity muscles, mainly the quadriceps and hamstrings. Although, open loop control strategies do not account for any changes in the muscles performance such as fatigue or load changes, they are widely used in clinics due to their relative simple implantation [1]. Closed loop controllers, as reported in many studies [2]–[4], use sensor feedback to update the stimulation levels (intensity and pulse width) as a response to any external disturbances. Some authors use a simple PID controller [5], Knee Extension Controller KEC [6], a combination of feedback and feedforward control or an adaptive approach [7]. Others use a first or a second order switching curve in the state space to control patient movements: The On/Off controller [3] and the ONZOFF controller [8], in the so-called “controller-centered” strategies. The main advantage of these strategies is their low number of parameters to be tuned during stimulation. The so-called “subject centered” strategies, (PDMR: Patient-Driven Motion Reinforcement [2], CHRELMS: Control by Handle REactions of Leg Muscle Stimulation [4]), introduce the voluntary contribution of the upper body of the patient as an essential part of the control diagram. This later is not yet adopted in clinical use because of

the relative high number of parameters to be identified. In order to overcome for the mentioned drawbacks, a better understanding of the muscle element as well as finding a compromise between a complex control strategy and a satisfactory one, should be taken into account. To accomplish this task, a new mathematical muscle model, has been developed [9], representing the complex physiological process. The number of recruited motor units increase as a function of both intensity stimulation I and Pulse Width PW . This phenomenon is modeled by an activation model (representing the ratio of recruited fibers α and the chemical control input u_{ch}) and a mechanical model (muscle contraction) (*fig.1*). The nonlinearities of the muscle model, and the required robustness regarding parameter variations and external disturbances lead us to adopt a controller relying on the sliding mode theory. The main issue concerns the muscle modeling and its use in the improvement of the human knee movement under FES by means of high order sliding mode controller (*HOSM*) [10]. In the next section, the system modeling is presented, it includes model of the knee-muscles and its state space formulation. In the third section, the theory of a high order sliding mode controller is presented. The results will be detailed in the fourth section.

II. SYSTEM MODELLING

A. Biomechanical model

The biomechanical model consists of two segments representing respectively the shank and thigh connected to each other by a revolute joint with one degree of freedom (*fig.2*). The thigh is supposed fixed with respect to the patient laying supine. Two agonist/antagonist muscles act

Fig. 1. The muscle model followed by the biomechanical model

Fig. 2. Biomechanical model of the knee with two muscles

on the knee: the quadriceps acts as an extensor muscle while the hamstrings is the flexor muscle. As a result two forces F_q and F_h cause respectively the extension and flexion of the knee. F_q and F_h are the inputs of the biomechanical model while the angle Θ is the corresponding output. ($\Theta = 0$ corresponds to full extension of the knee, $\Theta = 150^\circ$ corresponds to the maximum flexion and $\Theta = 90^\circ$ represents the resting position).

- L_0 = Thigh length,
- L_1 = Shank length,
- O = Center of rotation (Knee),
- L_{iq} = Distance between O and the insertion point of the quadriceps on the shank,
- L_{ii} = Distance between O and the insertion point of hamstrings on the shank,
- L_q = Length of the quadriceps,
- L_h = Initial length of the hamstrings,
- r = Pulley radius,
- H = Orthogonal projection of O on L_1 ,
- F_q = Extension force generated by the quadriceps,
- F_h = Flexion force generated by the hamstrings,
- G = Gravity force vector,
- m = Mass of the shank.

The geometric constraints allow us to evaluate quadriceps length L_q depending on the knee angle variable Θ :

$$L_q(\Theta) = \sqrt{L_0^2 - r^2} + r\Theta + \sqrt{L_{iq}^2 - r^2}$$

And the hamstrings length $L_h(\Theta)$:

$$L_h(\Theta) = \sqrt{L_0^2 + L_{ii}^2 + 2L_0L_{ii}\cos(\Theta)}$$

From the above equations, we can deduce the relative elongations of the quadriceps and the hamstrings:

$$\varepsilon_q(\Theta) = \frac{L_q - L_{0q}}{L_{0q}} = \frac{\sqrt{L_0^2 - r^2} + r\Theta + \sqrt{L_{iq}^2 - r^2} - L_{0q}}{L_{0q}}$$

$$\varepsilon_h(\Theta) = \frac{L_h - L_{0h}}{L_{0h}} = \frac{\sqrt{L_0^2 + L_{ii}^2 + 2L_0L_{ii}\cos(\Theta)} - L_{0h}}{L_{0h}}$$

L_{0q} and L_{0h} correspond respectively to the initial quadriceps and hamstrings lengths. Moment arm of the quadriceps is supposed to be constant and equal to the pulley radius while the moment arm of the hamstrings depends on the variable angle Θ .

$$\overline{OH} = \frac{L_0L_{ii}\sin(\Theta)}{\sqrt{L_0^2 + L_{ii}^2 + 2L_0L_{ii}\cos(\Theta)}}$$

From the above equations and the equation of motion which is a second order nonlinear dynamical equation, we obtained the acceleration $\ddot{\Theta}$ as a function of the inertia around the knee joint (I).

$$\ddot{\Theta} = \frac{1}{I} [r F_q - m g \cos(\Theta) \beta L_1 - F_v \dot{\Theta} - \frac{L_0L_{ii}\sin(\Theta)}{\sqrt{L_0^2 + L_{ii}^2 + 2L_0L_{ii}\cos(\Theta)}} F_h]$$

Parameters of the above model were taken for an average person: $L_0 = 50 \text{ cm}$, $L_{iq} = 4 \text{ cm}$, $L_{ii} = 5 \text{ cm}$, $r = 1 \text{ cm}$, $F_v = 0.5 \text{ N.m.s}$, $I = 0.0476 \text{ N.m.s}^2$, $g = 9.8 \text{ N.m}^{-2}$, $m = 3.5 \text{ Kg}$. β corresponds to the position of the center of gravity of the shank (< 1), F_v corresponds to coefficient of viscous friction.

B. Muscle model

In previous papers [9], [11], a new physiological skeletal muscle model has been proposed to describe the complex internal physiological phenomena. This model was developed in this current study to fulfill most of the physiological processes occurring during the phases of contraction and relaxation. We have adopted this model, in order to develop strategies for simulation, motion synthesis and motor control during clinical restoration of movement. Figure 3 shows the model with the parallel element E_p and two elements in series E_s (elastic element) and E_c (contractile element). This model is controlled by two variables: u_{ch} , a chemical control input and α , the ratio of the recruited fibers. We have described this model by two differential equations where the outputs are K_c and F_c representing, respectively, the stiffness and the force generated by the contractile element. K_0 and F_0 are the maximum values of K_c and F_c .

Fig. 3. Muscle model and particularization of EC

$$\begin{cases} \dot{K}_c = (s_0\alpha K_0 - s_u K_c + s_v q \frac{s_0\alpha F_0 K_c - s_u F_c K_c}{1+pK_c - s_v q F_c})u - \frac{s_v \alpha K_c}{1+pK_c - s_v q F_c} \\ \dot{F}_c = (\frac{s_0\alpha F_0 - s_u F_c}{1+pK_c - s_v q F_c})u + (\frac{bK_c - s_v \alpha F_c}{1+pK_c - s_v q F_c})\dot{\epsilon} \end{cases}$$

$$s_u = \text{sign}(u) = \begin{cases} -1 & \text{if } u < 0 \\ +1 & \text{if } u > 0 \end{cases}$$

$$s_v = \text{sign}(\dot{\epsilon}_c) = \begin{cases} +1 & \text{if } \dot{\epsilon}_c > 0 \\ -1 & \text{if } \dot{\epsilon}_c < 0 \end{cases}$$

$$s_0 = \frac{1+s_u}{2} \quad a = \frac{L_0}{L_{c0}} \quad b = L_0 \quad p = \frac{1}{K_s} \quad q = \frac{1}{L_{c0}K_s}$$

$$\epsilon_c = \frac{L_c - L_{c0}}{L_{c0}} \quad \epsilon_s = \frac{L_s - L_{s0}}{L_{s0}} \quad \epsilon = \frac{L - L_0}{L_0} \quad L = L_c + L_s$$

where s_u , s_0 and s_v are the signs of the control and the velocities of the contractile element, L_c and L_s represent respectively the length of the contractile and the elastic elements. The ratio of recruited fibers α is considered as a global scale factor which gives the percentage of the maximal possible force which can be generated by the muscle. Identification of the muscle parameters is ongoing and the simulations performed on the present version of the muscle model show satisfactory accuracy and prediction. The input signal for the muscle model represents the actual electrical signal as provided by the stimulator "PROSTIM" [12] offering the possibility of tuning three independent parameters: amplitude, pulse width and frequency. The parameters of the muscles were taken from [13] as follow:

Muscle model parameters	Variable	Numeric value (quad.-hamst.)	Unit
stiffness of E_s	K_s	1.10^4	N/m
Contractile element length E_c	L_{c0}	$41.10^{-2}, 38.10^{-2}$	m
Elastic element length E_s	L_{s0}	$8.10^{-2}, 10.10^{-2}$	m

C. Muscle-Knee: State space Model

The model of the muscles and knee joint can be rewritten as a non-linear state space function:

$$\dot{\mathbf{X}} = \mathbf{f}(\mathbf{x}, \mathbf{t}, \mathbf{U})$$

Where $\mathbf{X} = [X_1 \dots X_6]^T = [K_q \ K_h \ F_q \ F_h \ \Theta \ \dot{\Theta}]^T$ is the state vector and $\mathbf{U} = [u_q \ \alpha_q \ u_h \ \alpha_h]^T$ the control vector. The variable Θ represents the joint knee angle. The state variables K_q, F_q, u_q, α_q and K_h, F_h, u_h, α_h are respectively the state variables of the quadriceps and hamstrings. The state space model of the knee joint can be expressed as:

$$\dot{X}_1 = (s_{0q}\alpha_q K_{0q} + s_{vq}q q \frac{s_{0q}\alpha_q F_{0q} X_1 - s_{uq} X_3 X_1}{1+p_q X_1 - s_{vq} q q X_3}) u_q$$

$$- \frac{s_{vq}\alpha_q X_1 r X_6}{L_{0q} + p_q X_1 - s_{vq} q q X_3} - s_{uq} X_1 u_q$$

$$\dot{X}_2 = (s_{0h}\alpha_h K_{0h} + s_{vh}q h \frac{s_{0h}\alpha_h F_{0h} X_2 - s_{uh} X_4 X_2}{1+p_h X_2 - s_{vh} q h X_4}) u_h$$

$$- \frac{s_{vh}\alpha_h X_2 L_0 L_{ii} \sin(X_5)}{L_{0h} \sqrt{L_0^2 + L_{ii}^2} + 2L_0 L_{ii} \cos(X_5) (1+p_h X_2 - s_{vh} q h X_4)}$$

$$- s_{uh} X_2 u_h$$

$$\dot{X}_3 = \frac{s_{0q}\alpha_q F_{0q} - s_{uq} X_3}{1+p_q X_1 - s_{vq} q q X_3} u_q - \frac{b_q X_1 - s_{vq} \alpha_q X_3 r X_6}{L_{0q} (1+p_q X_1 - s_{vq} q q X_3)}$$

$$\dot{X}_4 = - \frac{b_h X_2 - s_{vh} \alpha_h X_4 L_0 L_{ii} \sin(X_5)}{L_{0h} \sqrt{L_0^2 + L_{ii}^2} + 2L_0 L_{ii} \cos(X_5) (1+p_h X_2 - s_{vh} q h X_4)}$$

$$+ \frac{s_{0h}\alpha_h F_{0h} - s_{uh} X_4}{1+p_h X_2 - s_{vh} q h X_4} u_h$$

$$\dot{X}_5 = X_6$$

$$\dot{X}_6 = \frac{1}{I} (X_3 r - X_4 \frac{L_0 L_{ii} \sin(X_5)}{\sqrt{L_0^2 + L_{ii}^2} + 2L_0 L_{ii} \cos(\Theta)}) - F_v X_6$$

$$- m g \cos X_5 \beta L_1)$$

III. SLIDING MODE CONTROLLER

The sliding mode control, has become recently widely used due to its high accuracy and robustness with respect to parameters uncertainty disturbances. The control task is to keep a constraint, given by equality of a smooth function called sliding surface, to zero. The dynamic smoothness in the vicinity of the sliding mode represents the sliding order of the system. As a generalisation of the classical sliding mode, this notion has been extended to the high order sliding mode. In this case, the control acts on the higher order time derivatives of the sliding variable instead of acting on its first time derivative. Thus the discontinuity of the control vector does not appear in the first $(r-1)^{th}$ total time derivative.

$$\frac{\partial s^{(i)}}{\partial u} = 0, (i = 1, 2, \dots, r-1), \frac{\partial s^{(r)}}{\partial u} \neq 0 \quad (1)$$

s , r represent respectively the sliding surface and the relative degree. u is the resulting control vector. Consequently we have:

$$s = \dot{s} = \ddot{s} = \dots = s^{r-1} = 0 \quad (2)$$

A. Position control law strategy

The sliding surface used to constraint the dynamic behavior of the knee joint is a first order differential equation chosen as:

$$s = (\dot{\Theta}_d - \dot{\Theta}) + \lambda(\Theta_d - \Theta) \quad (3)$$

Where $\dot{\Theta}_d$, Θ_d are respectively the desired velocity and desired position, λ is a positive coefficient. Higher values of λ , lead to a faster convergence along the sliding surface to the zero point of the phase-plane. Let us consider the sliding surface equation (3) in order to determine the relative order of the controlled system. We obtain the following result:

$$\frac{\partial \dot{s}}{\partial u} = 0, \frac{\partial \ddot{s}}{\partial u} \neq 0 \quad (4)$$

Therefore, the relative degree of the sliding mode control is $r = 2$. Considering the step response case ($\ddot{\Theta}_d = \dot{\Theta}_d = 0$), the second time order derivative of the sliding surface can be written as:

$$\ddot{s} = -\ddot{X}_6 - \lambda \dot{X}_6 \quad (5)$$

The expression of the second time derivative of the state variable X_6 is given by:

$$\ddot{X}_6 = \frac{1}{I} \left[\begin{aligned} & \frac{rs_{0q}\alpha_q F_{0q}}{1+p_q X_1 - s_{vq} q_q X_3} u_q - \frac{rs_{0q} X_3}{1+p_q X_1 - s_{vq} q_q X_3} u_q \\ & + r \frac{b_q X_1 - s_{vq} a_q X_3 r X_6}{L_{0q}(1+p_q X_1 - s_{vq} q_q X_3)} \\ & - \frac{s_{0h}\alpha_h F_{0h} L_0 L_{ii} \sin(X_5)}{(1+p_h X_2 - s_{vh} q_h X_4) \sqrt{L_0^2 + L_{ii}^2 + 2L_0 L_{ii} \cos(X_5)}} u_h \\ & + \frac{s_{0h} X_4 L_0 L_{ii} \sin(X_5)}{(1+p_h X_2 - s_{vh} q_h X_4) \sqrt{L_0^2 + L_{ii}^2 + 2L_0 L_{ii} \cos(X_5)}} u_h \\ & - \frac{(b_q X_1 - s_{vq} a_q X_3) r X_6 L_0 L_{ii} \sin(X_5)}{L_{0q}(1+p_q X_1 - s_{vq} q_q X_3) \sqrt{L_0^2 + L_{ii}^2 + 2L_0 L_{ii} \cos(X_5)}} \\ & - X_4 \frac{L_0 L_{ii} X_6 \cos(X_5) \sqrt{L_0^2 + L_{ii}^2 + 2L_0 L_{ii} \cos(X_5)}}{L_0^2 + L_{ii}^2 + 2L_0 L_{ii} \cos(X_5)} \\ & - X_4 \frac{L_0^2 L_{ii}^2 \sin^2(X_5)}{(L_0^2 + L_{ii}^2 + 2L_0 L_{ii} \cos(X_5)) \sqrt{L_0^2 + L_{ii}^2 + 2L_0 L_{ii} \cos(X_5)}} \\ & + mg\beta L_1 X_6 \sin(X_5) - F_v \dot{X}_6 \end{aligned} \right] \quad (6)$$

Inserting the expressions of $\dot{X}_6$ and $\ddot{X}_6$ within equation (5) allows writing the second time derivative of s as:

$$\ddot{s} = \varphi(x, t) + \gamma(t, x)u \quad (7)$$

It is assumed that $\Phi > 0$, $|\varphi| \leq \Phi$, $0 < \Gamma_m \leq \gamma \leq \Gamma_M$ [14], where s_0 , $u_0 < 1$, Γ_m , Γ_M and Φ are positive constants.

We express the equation (7) as:

$$\begin{cases} \dot{y}_1 = y_2 \\ \dot{y}_2 = \varphi(x, t) + \gamma(t, x)u \end{cases}$$

Where $y_1 = s$. In that case, the problem is equivalent to the finite time stabilization problem for the uncertain second-order system.

B. Statement of the control algorithm

[14] presented a range of 2-sliding algorithms to stabilise second order uncertain nonlinear systems. In the current study we have implemented the algorithm with prescribed law of variation of the sliding surface. This choice has been made based on criteria of robustness and finite time convergence [10]. The general formulation of such a class of a sliding mode control algorithm is:

$$\dot{u} = \begin{cases} -u & \text{if } |u| > 1 \\ -V_M \text{sign}(y_2 - g_c(y_1)) & \text{if } |u| \leq 1 \end{cases} \quad (8)$$

Where V_M is a positive constant and g_c a continuous function (Fig. 4). Moreover, this function must verify some specific conditions (see [10]).

$$g_c(y_1) = -\lambda_1 |y_1|^\rho \text{sign}(y_1), \quad \lambda_1 > 0, \quad 0.5 \leq \rho < 1 \quad (9)$$

The sufficient condition for the finite time convergence to the sliding manifold is defined by the following inequality:

$$V_M > \frac{\Phi + \sup[g_c(y_1)g_c(y_1)]}{\Gamma_m} \quad (10)$$

Larger values of λ_1 accelerate the convergence to reach the sliding surface and provide better robustness and stability. A substitution of y_2 by Δy_1 is theoretically possible whether y_2 is not available.

Fig. 4. Phase plot of the prescribed law convergence algorithm

IV. CO-CONTRACTION RESULTS

We have implemented the control algorithm defined by equation (8) on the simulator of the biomechanical model of the knee joint (*cf.* equation (6)). The components of the control vector U are the chemical inputs (u_q, u_h) and the ratio of the recruited fibers (α_q, α_h). These coefficients depend on the electrical stimulation current and Pulse Width modulation values. In our case the biomechanical model of the knee joint is controlled by two muscles: quadriceps and hamstrings. Consequently, there are two electrical currents, I_q and I_h as well as two Pulse Width Modulations values, (see [15]) PW_q and PW_h which have to be deduced from the control vector u . Muscle co-contraction can be defined as the simultaneous activation of agonist and antagonist muscle groups crossing the same joint and acting in the same plane [16]. The opposite muscles, quadriceps and hamstrings in this case, act simultaneously and thereby increase the stiffness at the knee joint. According to the state-space of the muscle-knee model (section IV), a co-contraction effect would increase $\dot{X}_3$ and $\dot{X}_4$, which will ensure a safer movement of the shank. Therefore, we propose a method to define the contribution of the control vector u stemming from the 2-sliding controller to calculate the needed electrical current stimulation values. According

Fig. 5. Co-contraction of the two muscles: quadriceps and hamstrings to the sign of the resulting control variable (u) at the

output of the *HOSM* Controller (*fig.5*), we have chosen to stimulate whether the quadriceps or the hamstrings. The co-contraction phenomenon was represented by a simultaneous contraction of the muscle and its antagonist muscle via a static ponderation ξ_q , ξ_h respectively for quadriceps and hamstrings.

$$If(u > 0) \Rightarrow \begin{cases} I_q &= \frac{u}{u_{nom}} I_{Max} \\ I_h &= \xi_q I_q \end{cases} \quad (11)$$

$$If(u < 0) \Rightarrow \begin{cases} I_q &= \xi_h I_h \\ I_h &= \frac{u}{u_{nom}} I_{Max} \end{cases} \quad (12)$$

u_{nom} and I_{Max} correspond respectively to nominal value of the control u and the maximal value authorized to stimulate a muscle (around 200 mA). The current values for quadriceps I_q and hamstrings I_h and the Pulse Width, respectively PW_q and PW_h , enable us to evaluate the required ratios of fibers to be recruited (α_q, α_h). The control vector u is defined as:

$$u = - \int V_M \text{sign}(y_2 - g_c(y_1)) dt \quad (if |u| \leq 1) \quad (13)$$

The chemical inputs u_q and u_h are automatically activated when the electrical currents are respectively superior to zero. We have implemented this algorithm on the simulator built with SimulinkTM software. We applied two different desired positions, starting from the rest position, $\Theta_d = 90^\circ$ as:

$$\begin{cases} 1) & 1s < t < 4s : \quad \Theta_d = 130^\circ \\ 2) & 6s < t < 9s : \quad \Theta_d = 50^\circ \\ 3) & Otherwise : \quad \Theta_d = 90^\circ \end{cases}$$

The coefficient of the 2-sliding controller were chosen to verify the condition equations (9). The following values have been used: $\lambda = 10$, $\lambda_1 = 20$, $\rho = 0.7$, $V_M = 1$. The simulation sampling periode was set to 10^{-4} sec. Figure 6(a) shows the step response for different desired angles. We notice in *Fig.6(b)* the finite time convergence of the sliding surface about 1sec in knee flexion and extension. Desired and current angle curves match when sliding surface reaches zero. In *Fig.7(a)*, we show the resulting simultaneous stimulation currents for quadriceps and hamstrings I_q and I_h . *Fig.7(b)* presents the control vector u computed by the equation (13).

If the resulting control variable u is positive (respectively negative) the quadricep (respectively hamstrings) is stimulated and controlled in a closed loop as shown in (*Fig.5*), while its antagonist muscle is controlled in an open loop control scaled by ξ_q (respectively ξ_h). The resulting control variable u was scaled, according to its sign by τ_1 (respectively τ_2) to the normalized amplitude of the current stimulation. Simulation results in *Fig.8* shows the alternative contraction of both quadriceps and hamstrings by setting $\xi_{q,h}$ to 0 forming thus, the On-Off control. *Fig.9* shows the sliding mode controller tracking a real joint angle trajectory recorded when walking at natural cadence [18]. The desired and the actual curves match very well. Parameters of the muscle model were identified by experiments on paraplegic patient *fig.10*.

Fig. 6. a) Desired step and actual knee angle variation, b) stabilization of the sliding surface

Fig. 7. a) Stimulation current (.. I_h , - I_q) $\xi_q = \xi_h$, b) The resulted control vector u

Fig. 8. a) Desired step and actual knee angle variation, b) Stimulation current (- I_h , .. I_q)

Fig. 9. Experimental and actual Knee angle variation recorded during a complete walking phase

Fig. 10. Experiments on paraplegic patient during knee angle and torque measurements were performed at the Centre Bouffard-Vercelli (Cerbère - France)

V. CONCLUSION

Few studies have treated the human muscle as an entire physiological element. Known by their robustness and accuracy, we used the sliding mode control. Because of the nonlinearity and the presence of a 2-order sliding system we adopted, in the current study, a high order sliding mode controller *HOSM*, which seems necessary to ensure a robust control and a safer movement of the lower extremities. This later was applied to a new multi-scale model developed within the *DEMAR* project. The muscle model is based on internal physiological characteristics assembling two levels: the microscopic one, involving the sliding actin-myosin filaments and the macroscopic part represented by a contractile element and an elastic element. We were able to control two antagonist muscles quadriceps and hamstrings alternatively and simultaneously (the so called co-contraction effect) with the same control vector, forcing dynamically the system to behave as a first order response. Satisfactory stability and tracking error were achieved after a finite time delay. The controller was able to track also a pre-recorded knee angle variation during a complete gait cycle. In order to validate simulation results further work will be carried out with paraplegic patients to evaluate the accuracy and the robustness of the high order sliding controller. Experiments are ongoing to validate the 2-sliding controller by using a multi-moment platform on a paraplegic patient. Other control strategies will be implemented as well to analyze the relevance of the muscle model in much complex situations.

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